Population scale mapping of transposable element diversity reveals links to gene regulation and epigenomic variation

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Abstract

Variation in the presence or absence of transposable elements (TEs) is a major source of genetic variation between individuals. Here, we identified 23,095 TE presence/absence variants between 216 wild Arabidopsis accessions. Most variants are rare, and we find a burden of rare variants associated with extremes of both gene expression and DNA methylation levels within the population. Of the common alleles identified, two thirds were not in linkage disequilibrium with nearby SNPs, implicating these TE variants as a source of novel genetic diversity. Nearly 200 common TE variants were associated with significantly altered expression of nearby genes, with potential functional consequences including decreased pathogen resistance. A major fraction of inter-accession DNA methylation differences were associated with nearby TE insertion variants, with DNA, indicating an important role of TE variants in facilitating epigenomic variation.

Introduction

Transposable elements (TEs) are mobile genetic elements present in nearly all studied organisms, and make up a large fraction of most eukaryotic genomes. The two classes of TEs are retrotransposons (class I elements), which transpose via an RNA intermediate requiring a reverse transcription reaction, and DNA transposons (class II elements), which transpose via either a cut-paste or, in the case of Helitrons, a rolling circle mechanism with no RNA intermediate [1]. TE activity poses mutagenic potential as a TE insertion may disrupt essential regions of the genome. Consequently, safeguards have evolved in order to suppress this activity. These safeguards include epigenetic transcriptional silencing mechanisms, chiefly involving the methylation of cytosine nucleotides (DNA methylation) to produce 5-methylcytosine (mC), a mark that can signal transcriptional silencing of the methylated locus. In *Arabidopsis thaliana* (Arabidopsis), DNA methylation occurs in three DNA sequence contexts: mCG, mCHG, and mCHH, where H is any base but G. Establishment of DNA methylation marks can be carried out by two distinct pathways – the RNA-directed DNA methylation pathway guided by 24 nucleotide (nt) small RNAs (smRNAs), and the DDM1/CMT2 pathway [2, 3]. A major function of DNA methylation in Arabidopsis is in the transcriptional silencing of TEs. Loss of DNA methylation due to mutations in genes essential for DNA methylation establishment and maintenance leads to expression of previously silent TEs, and in some cases transposition [2, 4–8].

TEs are thought to play an important role in evolution, not only because of the disruptive potential of their transposition. The release of transcriptional and post-transcriptional silencing of TEs can lead to bursts of TE activity, quickly generating new genetic diversity [9]. TEs may carry regulatory information such as promoters and transcription factor binding sites, and their mobilization may lead to the creation or expansion of gene regulatory networks [10–13]. Furthermore, the transposase enzymes required and encoded by TEs have frequently been domesticated and repurposed as endogenous proteins, such as the *DAYSLEEPER* gene in Arabidopsis, derived from a hAT transposase enzyme [14]. Clearly, the activity of TEs can have widespread and unpredictable effects on the host genome. However, the identification of TE presence/absence variants in genomes has remained difficult to date. It is challenging to identify the structural changes in the genome caused by TE mobilization using current short-read sequencing technologies as these reads are typically mapped to a reference genome, which has the effect of masking structural changes that may be present. However, in terms of the number of base pairs affected, a large fraction of genetic differences between Arabidopsis accessions appear to be due to variation in TE content [15, 16]. Therefore identification of TE variants is essential in order to develop a more comprehensive understanding of the genetic variation that exists between genomes, and of the consequences of TE movement on genome and cellular function.

The tools developed previously for identification of novel TE insertion sites have several limitations. They either require a library of active TE sequences, cannot identify TE absence variants, or are not designed with population studies in mind, or suffer from a high degree of false-negatives [16–19]. In order to accurately map the locations of TE presence/absence variants with respect to a reference genome, we have developed a novel algorithm, TEPID (Transposable Element Polymorphism IDentification), which is designed for population studies. We tested our algorithm using both simulated and real Arabidopsis sequencing data, finding that TEPID is able to accurately identify TE presence/absence variants with respect to the Col-0 reference genome. We applied our TE analysis method to existing genome resequencing data for 216 different wild Arabidopsis accessions, and identified widespread TE variation amongst these accessions [20].

Results

Computational identification of TE presence/absence variation

We developed TEPID, an analysis pipeline capable of detecting TE presence/absence variants from paired end DNA sequencing data. TEPID integrates split and discordant read mapping information, read mapping quality, sequencing breakpoints, as well as local variations in sequencing coverage to identify novel TE presence/absence variants with respect to a reference TE annotation (Figure 1; see methods). This typically takes 5-10 minutes using Arabidopsis sequencing data at 20-40x coverage (excluding the read mapping step). After TE variant discovery has been performed, TEPID then includes a second refinement step designed for population studies. This examines each region of the genome where there was a TE insertion identified in any member of the group, and checks for evidence of this insertion in all other members of the group. In this way, TEPID leverages TE variant information for a group of related samples to correct false negative calls within the group. Testing of TEPID using simulated TE variants in the Arabidopsis genome showed that it was able to reliably detect simulated TE variants at sequencing coverage levels commonly used in genomics studies (Figure 1 - figure supplement 1).

In order to further assess the sensitivity and specificity of TE variant discovery using TEPID, we identified TE variants in the *Landsberg erecta* (L*er*) accession, and compared these with the L*er* genome assembly created using long PacBio sequencing reads [21]. Previously published 100 bp paired-end L*er* genome resequencing reads [22] were first analyzed using TEPID, enabling identification of 446 TE insertions (Figure 1 - source data 1) and 758 TE absence variants (Figure 1 - source data 2) with respect to the Col-0 reference TE annotation. Reads providing evidence for these variants were then mapped to the L*er* reference genome that was generated by *de novo* assembly using Pacific Biosciences P5-C3 chemistry with a 20 kb insert library [21], using the same alignment parameters as was used to map reads to the Col-0 reference genome. This resulted in 98.7% of reads being aligned concordantly to the L*er* reference, whereas 100% aligned discordantly or as split reads to the Col-0 reference genome (Table 1). To find whether reads mapped to homologous regions in both the Col-0 and L*er* reference genomes, we conducted a blast search [23] using the DNA sequence between read pair mapping locations in the L*er* genome against the Col-0 genome, and found the top blast result for 80% of reads providing evidence for TE insertions, and 89% of reads providing evidence for TE absence variants in L*er*, to be located within 200 bp of the TE variant reported by TEPID. We conclude that reads providing evidence for TE variants map discordantly or as split reads when mapped to the Col-0 reference genome, but map concordantly to homologous regions of the L*er* *de novo* assembled reference genome, indicating that structural variation is present at the sites identified by TEPID, and that this is resolved in the *de novo* assembled genome.

To estimate the rate of false negative TE absence calls made using TEPID, we compared our L*er* TE absence calls to the set of TE absences in L*er* genome identified previously by aligning full-length Col-0 TEs to the Ler reference using BLAT [16]. We found that 89.6% (173/193) of these TE absences were also identified using TEPID, indicating a false negative rate of around 10% for TE absence calls. To determine the rate of false negative TE insertion calls, we ran TEPID using 90 bp paired-end Col-0 reads (Col-0 control samples from [24]), aligning reads to the L*er* PacBio assembly. As TEPID requires a high-quality TE annotation to discover TE variants, which is not available for the L*er* assembly, we simply looked for discordant and split read evidence at the known Col-0-specific TE insertion sites [16], and found evidence reaching the TEPID threshold for a TE insertion call to be

made at 89.6% (173/193) of these sites, indicating a false negative rate of 10%. However, we note that this estimate does not take into account the TEPID refinement step used on large populations, and so the false negative rate for samples analyzed in the population from Schmitz et al. (2013) is likely to be slightly lower than this estimate, as each accession gained on average 4% more insertion calls following this refinement step (Figure 2 - figure supplement 1).

Abundant TE positional variation among natural Arabidopsis populations

We used TEPID to analyze previously published 100 bp paired-end genome resequencing data for 216 different Arabidopsis accessions [20], and identified 15,007 TE insertions and 8,088 TE absence variants relative to the Col-0 reference accession, totalling 23,095 unique TE variants. In most accessions we identified 300-500 TE insertions (mean = 378) and 1,000-1,500 TE absence variants (mean = 1,279), the majority of which were shared by two or more accessions (Figure 2 - figure supplement 2). PCR validations were performed for a random subset of 10 insertions and 10 absence variants in 14 accessions (totalling 280 validations), and confirmed the high accuracy of TE variant discovery using the TEPID package, with results similar to that observed using simulated data and the L*er* genome analysis (Figure 2 - figure supplement 3). The number of TE insertions identified was positively correlated with sequencing depth of coverage, while the number of TE absence variants identified had no correlation with sequencing coverage (Figure 2 - figure supplement 4A, B), indicating that the sensitivity of TE absence calls is not limited by sequencing depth, while TE insertion calls benefitted from high sequencing depth. However, accessions with low coverage gained more TE insertion calls during the TEPID refinement step (Figure 2 - figure supplement 4C), indicating that these false negatives were effectively reduced by leveraging TE variant information for the whole population.

As TE insertion and TE absence calls represent an arbitrary comparison to the Col-0 reference genome, we sought to remove these arbitrary comparisons and classify each variant as a new TE insertion or true deletion of an ancestral TE in the population. To do this, we examined the minor allele frequency (MAF) of each variant in the population, under the expectation that the minor allele is the derived allele. We re-classified common TE absences relative to Col-0 as rare TE insertions in Col-0, and common TE insertions relative to Col-0 as true TE deletions in Col-0. Cases where the TE variant was at an high MAF (>20%) were assigned NA calls, as it could not be determined if these were cases where the variant was most likely to be due to a true TE deletion or due to a new TE insertion. While these classifications are not definitive, as there will be rare cases where a true TE deletion has spread through the population and becomes the common allele, we will correctly classify most TE variants. Overall, we found 72.3% of the TE absence variants identified with respect to the Col-0 reference genome were likely due to a true TE deletion in these accessions, while 4.8% were due to new insertions in Col-0 not shared by others in the population (Table 2). Overall, we identified 15,077 new TE insertions, 5,856 true TE deletions, and 2,162 TE variants at a high MAF that were unable to be classified as an insertion or deletion.

TE insertions and deletions were distributed throughout chromosome 1 in a pattern that was similar to the distribution of all Col-0 TEs (Figure 2A). TE deletions and common TE variants were found in similar chromosomal regions, as deletion variants represent the rare loss of common variants. TE deletions and common variants were more highly enriched in the pericentromeric regions than rare variants or TE insertions. The distribution of rare TE variants and TE insertions was somewhat similar to that observed for regions of the genome previously identified as being differentially methylated

in all DNA methylation contexts (mCG, mCHG, mCHH) between the wild accessions (population C-DMRs), while population CG-DMRs, differentially methylated in the mCG context, less frequently overlapped with all types of TE variants identified [20]. Furthermore, TE variants were depleted within genes and DNase I hypersensitivity sites [25], while they were enriched in gene flanking regions and within other annotated TEs or pseudogenes (Figure 2B). We found that TE deletions and common TE variants were enriched within the set of TE variants found in gene bodies (Figure 2C, D). We did not find any significant enrichment of TE variants within the *KNOT ENGAGED ELEMENT (KEE)* regions, previously identified as regions that may act as a "TE sink" [26] (Figure 2 - figure supplement 5). This may indicate that these regions do not act as a "TE sink" as has been previously proposed, or that the "TE sink" activity is restricted to very recent insertions, as the insertions we analysed in this population were likely older than those used in the study by Grob et al.

Among the identified TE variants, several TE superfamilies were over- or under-represented compared to the number expected by chance given the overall genomic frequency of different TE types (Figure 2E). In particular, both TE insertions and deletions in the RC/Helitron superfamily were less numerous than expected, with an 11.5% depletion of RC/Helitron elements in the set of TE variants. In contrast, TEs belonging to the LTR/Gypsy superfamily were more frequently deleted than expected, with a 17% enrichment in the set of TE deletions. This was unlikely to be due to a differing ability of our detection methods to identify TE variants of different lengths, as the TE variants identified had a similar distribution of lengths as all Arabidopsis TEs annotated in the Col-0 reference genome (Figure 2 - figure supplement 6). These enrichments suggest that the RC/Helitron TEs have been relatively dormant in recent evolutionary history, while the LTR/Gypsy, which are highly enriched in the pericentromeric regions, are frequently lost from the Arabidopsis genome. At the family level, we observed similar patterns of TE variant enrichment or depletion (Figure 2 - figure supplement 7; source data 5).

We further examined Arabidopsis (Col-0) DNA sequencing data from a transgenerational stress experiment to investigate the possible minimum number of generations required for TE variants to arise [24]. We identified a single potential TE insertion in a sample following 10 generations of single-seed descent under high salinity stress conditions, and no TE variants in the control single-seed descent set. However, without experimental validation it remains unclear if this represents a true variant. Therefore, we conclude that TE variants may arise at a rate less than 1 insertion in 30 generations under laboratory conditions. Further experimental work will be required to precisely determine the rate of transposition in Arabidopsis.

Relationship between TE variants and single nucleotide polymorphisms

Although thousands of TE variants were identified, they may be linked (i.e. 'tagged') by the previously identified single nucleotide polymorphisms (SNPs), or unlinked from SNPs across the accessions. We tested how often common TE variants (>3 % MAF; 7 accessions) were linked to adjacent SNPs to determine when they would represent a previously unassessed source of genetic variation between accessions. SNPs that were previously identified between the accessions [20] were compared to the presence/absence of individual TE variants. For the testable TE variants in the population, the nearest 300 flanking SNPs upstream and downstream of the TE variant site were analyzed for local linkage disequilibrium (LD, r^2; see methods). TE variants were classified as being either 'low', 'mid', or 'high' LD variants by comparing ranked r^2 values of TE variant to SNPs against the median ranked r^2 value for all between SNP comparisons (SNP-SNP) to account for regional variation in the extent

of SNP-SNP LD (Figure 3A, B) due to recombination rate variation or selection [27]. The majority (61%) of testable TE variants had low LD with nearby SNPs, and represent a source of genetic diversity not previously assessed by current SNP-based genotype calling methods (Figure 3C). 29% of TE variants displayed high levels of LD and are tagged by nearby SNPs, while only 10% had intermediate levels of LD. We observed a positive correlation between TE variant MAF and LD state, with variants of a high minor allele frequency more often classified as high-LD (Figure 3D). While the proportion of TE variants classified as high, mid, or low-LD was mostly the same for both TE insertions and TE deletions, TE variants with a high MAF (>20%) that were unable to be classified as either true deletions or as new insertions had a much higher proportion of high-LD variants (Figure 3E). This was consistent with the observation that the more common alleles were more often in a high-LD state. TE variants displayed a similar distribution over chromosome one regardless of linkage classification (Figure 3 - figure supplement 1). Overall, this analysis revealed an abundance of previously unexplored genetic variation that exists amongst Arabidopsis accessions caused by the presence or absence of TEs, and illustrates the importance of identifying TE variants alongside other genetic diversity such as SNPs.

TE variants affect gene expression

To determine whether the TE variants identified affected nearby gene expression, we compared the steady state transcript abundance within mature leaf tissue between accessions with and without TE insertions or deletions, for genes with TE variants located in the 2 kb gene upstream region, 5' UTR, exon, intron, 3' UTR or 2 kb downstream region (Figure 4A). While the steady state transcript abundance of most genes appeared to be unaffected by the presence of a TE, 196 genes displayed significant differences in transcript abundance linked with the presence of a TE variant, indicating a role for these variants in the local regulation of gene expression (1% false discovery rate; greater than 2-fold change in transcript abundance; Figure 4A, Figure 4 - source data 1). We did not find any functional category enrichments in this set of differentially expressed genes. As rare TE variants with a MAF less than 3% may also be associated with a difference in transcript abundance, but were unable to be statistically tested due to their rarity, we performed a burden test for enrichment of rare variants in the extremes of expression [28]. Briefly, this method counts the frequency of rare variants within each gene expression rank in the population, and aggregates this information over the entire population to determine whether an enrichment of rare variants exists within any gene expression rank. We found a strong enrichment for gene expression extremes for TE variants in the all gene features tested (Figure 4B). While TE variants in gene upstream regions showed a strong enrichment of both high and low gene expression ranks, TE variants in exons or gene downstream regions seemed to have a stronger enrichment for low expression ranks than high ranks. Randomization of the accession names removed these enrichments completely (Figure 4 - figure supplement 1), and there was little difference between TE insertions and TE deletions in the gene expression rank enrichments found (Figure 4 - figure supplement 2). This rare variant analysis further indicated that TE variants may alter the transcript abundance of nearby genes.

As both increases and decreases in transcript abundance of nearby genes were observed for TE variants within each gene feature, it appears to be difficult to predict the impact a TE variant may have on nearby gene expression. Furthermore, gene-level transcript abundance measurements may fail to identify the potential positional effect of TE variants upon transcription. To more closely examine changes in transcript abundance associated with TE variants among the accessions, we

inspected a subset of TE variant sites and identified TE variants that appear to have an impact on transcriptional patterns beyond changes in total transcript abundance from a nearby gene. For example, the presence of a TE insertion within an exon of *AtRLP18* (AT2G15040) was associated with truncation of the transcripts at the TE insertion site in accessions possessing the TE variant, as well as silencing of a downstream gene encoding a leucine-rich repeat protein (AT2G15042) (Figure 5A, B). Both genes had significantly lower transcript abundance in accessions containing the TE insertion ($p < 5.8 \times 10^{-10}$, Mann-Whitney U test). *AtRLP18* is reported to be involved in bacterial resistance, with the disruption of this gene by T-DNA insertion mediated mutagenesis resulting in increased susceptibility to the bacterial plant pathogen *Pseudomonas syringae* [29]. We examined pathogen resistance phenotype data [30], and found that accessions containing the TE insertion in the *AtRLP18* exon were more often sensitive to infection by *Pseudomonas syringae* transformed with *avrPpH3* genes (Figure 5C). This suggests that the accessions containing this TE insertion within *AtRLP18* may have an increased susceptibility to certain bacterial pathogens.

We also observed some TE variants associated with increased expression of nearby genes. For example, the presence of a TE within the upstream region of a gene encoding a pentatricopeptide repeat (PPR) protein (AT2G01360) was associated with higher steady state transcript abundance of this gene (Figure 5D, E). Transcription appeared to begin at the TE insertion point, rather than the transcriptional start site of the gene (Figure 5D). Accessions containing the TE insertion had significantly higher AT2G01360 transcript abundance than the accessions without the TE insertion ($p < 1.8 \times 10^{-7}$, Mann-Whitney U test). The apparent transcriptional activation, linked with presence of a TE belonging to the *HELITRON1* family, indicates that this element may carry sequences or other regulatory information that has altered the expression of genes downstream of the TE insertion site. Importantly, this variant was classified as a low-LD TE insertion, as it is not in LD with surrounding SNPs, and therefore the associated changes in gene transcript abundance would not be identified using only SNP data. This TE variant was also upstream of *QPT* (AT2G01350), involved in NAD biosynthesis [31], which did not show alterations in steady state transcript abundance associated with the presence of the TE insertion, indicating a potential directionality of regulatory elements carried by the TE (Figure 5D, E). Overall, these examples demonstrate that TE variants can have unpredictable, yet important, effects on the expression of nearby genes, and these effects may be missed by studies focused on genetic variation at the level of SNPs.

TE variants explain many DNA methylation differences between accessions

As TEs are frequently highly methylated in Arabidopsis [32–35], we next assessed the DNA methylation state surrounding TE variant sites to determine whether TE variants might be responsible for some of the differences in DNA methylation patterns previously observed between the wild accessions [20]. We found that TE variants were often physically close to DMRs (Figure 6A). Furthermore, C-DMRs were more often close to a TE insertion than expected, while they were rarely near a TE deletion (Table 3). CG-DMRs were rarely close to TE insertions or TE deletions. Overall, we found 48% of the 13,482 previously reported population C-DMRs were located within 1 kb of a TE variant (predominantly TE insertions), while only 15% of CG-DMRs were within 1 kb of a TE variant (Table 3). For C-DMRs, this was significantly more than expected by chance, while it was significantly less than expected for CG-DMRs ($p < 1 \times 10^{-4}$, determined by resampling 10,000 times). To determine if DMR methylation levels were dependent on the presence/absence of nearby TE variants, we calculated Pearson correlation coefficients between the DNA methylation level at each DMR and

the presence/absence of the nearest TE variant. We observed a negative correlation between the distance from a C-DMR to the nearest TE insertion and the correlation between the DNA methylation level at the C-DMR with the presence/absence of the TE insertion (Figure 6B). This suggests a distance-dependent effect of TE insertion presence on C-DMR methylation. In contrast, we found no such relationship for TE deletions on C-DMRs, or for insertions or deletions on CG-DMRs (Figure 6B). DNA methylation levels at C-DMRs located within 1 kb of a TE insertion (TE-DMRs) were more often positively correlated with the presence/absence of a TE insertion than the DNA methylation levels at C-DMRs further than 1 kb from a TE insertion (non-TE-DMRs). This was evident from the distribution of correlations for non-TE-DMRs being centred around zero, whereas for TE-DMRs this distribution was skewed to the right (Figure 6C, D=0.23). For TE deletions, we did not observe such a difference in the distributions of correlation coefficients between TE-DMRs and non-TE-DMRs, nor for CG-DMRs and their nearby TE insertions or deletions (Figure 6C, D=0.07-0.10). Furthermore, DNA methylation levels were often higher in the presence of the nearby TE insertion, while this relationship was generally not observed for C-DMRs further than 1 kb from a TE variant, for TE deletions, or for CG-DMRs (Figure 6 - figure supplement 1).

As the above correlations between TE presence/absence and DMR methylation level rely on the TE variants having a high MAF, this precludes an analysis of the effect of rare variants on DMR methylation levels. To determine the effect that these rare TE variants may have on DMR methylation levels, we performed a burden test for enrichment of DMR methylation extremes at TE-DMRs, similar to as was done to test the effect of rare variants on gene expression. We found a strong enrichment for high C-DMR and CG-DMR methylation level ranks for TE insertions, while TE deletions were associated with both high and low extremes of DNA methylation levels at C-DMRs, and less so at CG-DMRs (Figure 6D). This further indicates that the presence of a TE insertion is associated with higher C-DMR methylation levels, while TE deletions seem to have more variable effects on DMR methylation levels. This enrichment was completely absent after repeating the analysis with randomized accession names (Figure 6 - figure supplement 2). We also observed a slight enrichment for low DMR methylation ranks for TE insertions near CG-DMRs, indicating that the insertion of a TE was sometimes associated with reduced CG methylation in nearby regions (<1 kb from the TE). We examined these TE insertions in a genome browser, and found that some TE insertions were associated with decreased transcript abundance of nearby genes, with a corresponding loss of gene body methylation, offering a potential explanation for the decreased CG methylation observed near some TE insertions (Figure 6 - figure supplement 3).

We next examined levels of DNA methylation in regions flanking all TE variants regardless of the presence or absence of a population DMR call. While DNA methylation levels around pericentromeric TE insertions and deletions (<3 Mb from a centromere) seemed to be unaffected by the presence of a TE insertion (Figure 7A), TE insertions in the chromosome arms were associated with an increase in DNA methylation levels in all contexts (Figure 7A, B). In contrast, TE deletions in the chromosome arms did not affect patterns of DNA methylation, as the flanking methylation level in all contexts seemed to remain high following deletion of the TE (Figure 7A, C). As the change in DNA methylation levels around TE variant sites appeared to be restricted to regions <200 bp from the insertion site, we correlated DNA methylation levels in 200 bp regions flanking TE variants with the presence/absence of TE the variants. DNA methylation levels were often positively correlated with the presence of a TE insertion when the insertion was distant from a centromere (Figure 7D). TE deletions were more variably correlated with local DNA methylation levels, but also showed a slight bias towards positive correlations for TE deletions distant from the centromeres. As methylome data was available for both leaf and bud tissue for 12 accessions, we repeated this analysis comparing between tissue types, but

did not observe any difference in the patterns of methylation surrounding TE variant sites between the two tissues (Figure 7 - figure supplement 1).

These results indicate that local DNA methylation patterns are influenced by the differential TE content between genomes, and that the DNA methylation-dependent silencing of TEs may lead to formation of DMRs between wild Arabidopsis accessions. TE insertions appear to be important in defining local patterns of DNA methylation, while DNA methylation levels often remain elevated following a TE deletion, and so are independent from the presence or absence of TEs in these cases. Importantly, the distance from a TE insertion to the centromere appears to have a strong impact on whether an alteration of local DNA methylation patterns will occur. This is likely due to flanking sequences being highly methylated in the pericentromeric regions, and so the insertion of a TE cannot further increase levels of DNA methylation. Overall, a large fraction of the population C-DMRs previously identified between wild accessions are correlated with the presence of local TE insertions, but not TE deletions. CG-DMR methylation levels seem to be mostly independent from the presence/absence of common TEs variants, while rare TE variants have an impact on DNA methylation levels at both C-DMRs and CG-DMRs.

Genome-wide analysis of TE variant association with methylation levels highlights distant and local control

To better quantify the above results, an association scan was conducted for all common TE variants (>3% MAF) and all population C-DMRs for the 124 accessions with both DNA methylation and TE variant data available. To test the significance of each pairwise correlation, we collected bootstrap p-value estimates based on 500 permutations of accession labels. TE-DMR associations were deemed significant if they had a real association more extreme than any of the 500 permutations ($p < 1/500$). A band of significant associations was observed for TE insertions and their nearby C-DMRs, signifying a local association between TE insertion presence/absence and C-DMR methylation (Figure 8A). This local association was not as strong for TE deletions (Figure 8B), consistent with our above findings. While TE variants and DNA methylation showed a local association, it is also possible that TE variation may influence DNA methylation state more broadly in the genome, perhaps through production of *trans*-acting smRNAs or inactivation of genes involved in DNA methylation establishment or maintenance. To identify any potential enrichment of C-DMRs regulated in *trans*, we summed the total number of significant associations for each TE variant across the whole genome (Figure 8A and B, top panels). At many sites, we found far more significant associations than expected due to the false positive rate alone. This suggested the existence of many putative *trans* associations between TE variants and genome-wide C-DMR methylation levels. We further examined these C-DMRs that appeared to be associated with a TE insertion in *trans*, checking for TE insertions near these C-DMRs that were present in the same accessions as the *trans* associated TE, as these could lead to a false *trans* association. These were extremely rare, with only 4 such cases for TE insertions, and 38 cases for TE deletions, and so were unable to explain the high degree of *trans* associations found.

Discussion

Here we discovered widespread differential TE content between wild Arabidopsis accessions, and explore the impact of these variants at the level of individual accessions. Most TE variants were due to the *de novo* insertion of TEs, while a smaller subset was likely due to the deletion of ancestral TE copies, mostly around the pericentromeric regions. A subset (32%) of TE variants with a minor allele frequency above 3% were able to be tested for linkage with nearby SNPs. The majority of these TE variants were not in LD with SNPs, indicating that they represent genetic variants currently overlooked in genomic studies. We found a marked depletion of TE variants within gene bodies and DNase I hypersensitivity sites (putative regulatory regions), indicating that the more deleterious TE insertions may have been removed from this population through selection. Of those TE variants found in gene bodies, TE deletions were overrepresented, indicating that the loss of ancestral TEs inserted within genes may be more frequent, or perhaps less deleterious, than the *de novo* insertion of TEs into genes.

The identification of a large number of TE variants in this population gave an opportunity to form statistically robust correlations between TE presence/absence and transcript abundance from nearby genes, as well as genome-wide patterns of DNA methylation. We were able to identify examples where TE variants appear to have an effect upon gene expression, both in the disruption of transcription and in the spreading or disruption of regulatory information leading to the transcriptional activation of genes, indicating that these TE variants can have important consequences upon the expression of protein coding genes (Figure 5). In one case, these changes in gene expression could be linked with phenotypic changes, with accessions containing a TE insertion more frequently sensitive to bacterial infection. Further experiments will be needed to establish a causal link between this TE insertion and the associated phenotype. An analysis of rare TE variants, present at a low MAF, further strengthened this relationship between TE presence/absence and altered transcript abundance, as we were able to identify a strong enrichment of rare TE variants in accessions with extreme gene expression ranks in the population.

Perhaps most importantly, we provide evidence that differential TE content between genomes of Arabidopsis accessions underlies a large fraction of the previously reported population C-DMRs, in agreement with recent similar findings [16]. Thus, the frequency of pure epialleles, independent of underlying genetic variation, may be even more rare than previously anticipated [36]. We did not find evidence of CG-DMR methylation level being altered by the presence of common TE variants, but rather rare TE variants may be more important in shaping patterns of DNA methylation at some CG-DMRs, though the reasons for this distinction remain unclear. The level of local DNA methylation changes associated with TE variants was also related to the distance from a TE variant to the centromere, with variants in the chromosome arms being more strongly correlated with DNA methylation levels. This seems to be due to a higher baseline level of DNA methylation at the pericentromeric regions, which prevent any further increase in DNA methylation level following insertion of a TE. Furthermore, we found an important distinction between TE insertions and TE deletions in the effect that these variants have on nearby DNA methylation levels. While flanking DNA methylation levels appeared to increase following a TE insertion, the deletion of an ancestral TE was often not associated with a corresponding decrease in flanking DNA methylation levels (Figure 7). This indicates that high levels of DNA methylation, once established, may be maintained in the absence of the TE insertion that presumably triggered the original change in DNA methylation level. It is then possible that TE variants explain more of the variation in DNA methylation patterns than we

find direct evidence for, if some C-DMRs were formed by the insertion of an ancestral TE that is now absent in all the accessions analysed here. These DMRs would then represent the epigenetic "scars" of past TE insertions.

Finally, we performed a genome-wide scan of common TE variant association with C-DMR methylation levels, and found further evidence of a strong local association between TE insertion presence/absence and C-DMR methylation level (Figure 8). We were also able to identify some TE variants that appeared to be associated with changes in DNA methylation levels at multiple loci throughout the genome, indicating a possible *trans* regulation of DNA methylation levels linked to certain TE variants. Further experiments will be required to confirm and examine the role of these TE variants in determining genome-wide patterns of DNA methylation. Overall our results show that TE presence/absence variants between wild Arabidopsis accessions not only have important effects on nearby gene expression, but can also have a role in determining local patterns of DNA methylation, and explain many regions of differential DNA methylation previously observed in the population.

Methods

TEPID development

Mapping

FASTQ files are mapped to the reference genome using the 'tepid-map' algorithm (Figure 1). This first calls bowtie2 [37] with the following options: '–local', '–dovetail', '–fr', '-R5', '-N1'. Soft-clipped and unmapped reads are extracted using Samblaster [38], and remapped using the split read mapper Yaha [39], with the following options: '-L 11', '-H 2000', '-M 15', '-osh'. Split reads are extracted from the Yaha alignment using Samblaster [38]. Alignments are then converted to bam format, sorted, and indexed using samtools [40].

TE variant discovery

The 'tepid-discover' algorithm examines mapped bam files generated by the 'tepid-map' step to identify TE presence/absence variants with respect to the reference genome. Firstly, mean sequencing coverage, mean library insert size, and standard deviation of the library insert size is estimated. Discordant read pairs are then extracted, defined as mate pairs that map more than 4 standard deviations from the mean insert size from one another, or on separate chromosomes.

To identify TE insertions with respect to the reference genome, split read alignments are first filtered to remove reads where the distance between split mapping loci is less than 5 kb, to remove split reads due to small indels, or split reads with a mapping quality (MAPQ) less than 5. Split and discordant read mapping coordinates are then intersected using pybedtools [41, 42] with the Col-0 reference TE annotation, requiring 80% overlap between TE and read mapping coordinates. To determine putative TE insertion sites, regions are then identified that contain independent discordant read pairs aligned in an orientation facing one another at the insertion site, with their mate pairs intersecting with the same TE (Figure 1). The total number of split and discordant reads intersecting the insertion site and the TE is then calculated, and a TE insertion predicted where the combined number of reads is greater than a threshold determined by the average sequencing depth over the whole genome (1/10 coverage if coverage is greater than 10, otherwise a minimum of 2 reads). Alternatively, in the

absence of discordant reads mapped in orientations facing one another, the required total number of split and discordant reads at the insertion site linked to the inserted TE is set higher, requiring twice as many reads.

To identify TE absence variants with respect to the reference genome, split and discordant reads separated >20 kb from one another are first removed, as 99.9% of Arabidopsis TEs are shorter than 20 kb, and this removes split reads due to larger structural variants not related to TE diversity. Col-0 reference annotation TEs that are located within the genomic region spanned by the split and discordant reads are then identified. TE absence variants are predicted where at least 80% of the TE sequence is spanned by a split or discordant read, and the sequencing depth within the spanned region is <10% the sequencing depth of the 2 kb flanking sequence, and there are a minimum number of split and discordant reads present, determined by the sequencing depth (1/10 coverage; Figure 1). A threshold of 80% TE sequence spanned by split or discordant reads is used, as opposed to 100%, to account for misannotation of TE sequence boundaries in the Col-0 reference TE annotation, as well as TE fragments left behind by DNA TEs during cut-paste transposition (TE footprints) that may affect the mapping of reads around annotated TE borders [43]. This was found to improve TE absence detection using simulated data. Furthermore, the coverage within the spanned region may be more than 10% that of the flanking sequence, but in such cases twice as many split and discordant reads are required. If multiple TEs are spanned by the split and discordant reads, and the above requirements are met, multiple TEs in the same region can be identified as absent with respect to the reference genome. Absence variants in non-Col-0 accessions are subsequently recategorized as TE insertions present in the Col-0 genome but absent from a given wild accession.

TE variant refinement

Once TE insertions are identified using the 'tepid-map' and 'tepid-discover' algorithms, these variants can be refined if multiple related samples are analysed. The 'tepid-refine' algorithm is designed to interrogate regions of the genome in which a TE insertion was discovered in other samples but not the sample in question, and check for evidence of that TE insertion in the sample using lower read count thresholds compared to the 'tepid-discover' step. In this way, the refine step leverages TE variant information for a group of related samples to reduce false negative calls within the group. This distinguishes TEPID from other similar methods for TE variant discovery utilizing short sequencing reads. A file containing the coordinates of each insertion, and a list of sample names containing the TE insertion must be provided to the 'tepid-refine' algorithm, which this can be generated using the 'merge_insertions.py' script included in the TEPID package. Each sample is examined in regions where there was a TE insertion identified in another sample in the group. If there is a sequencing breakpoint within this region (no continuous read coverage spanning the region), split reads mapped to this region will be extracted from the alignment file and their coordinates intersected with the TE reference annotation. If there are split reads present at the variant site that are linked to the same TE as was identified as an insertion at that location, this TE insertion is recorded in a new file as being present in the sample in question. If there is no sequencing coverage in the queried region for a sample, an "NA" call is made indicating that it is unknown whether the particular sample contains the TE insertion or not.

While the above description relates specifically to use of TEPID for identification of TE variants in Arabidopsis in this study, this method can be also applied to other species, with the only prerequisite being the annotation of TEs in a reference genome and the availability of paired-end DNA sequencing data.

TE variant simulation

To test the sensitivity and specificity of TEPID, 100 TE insertions (50 copy-paste transpositions, 50 cut-paste transpositions) and 100 TE absence variants were simulated in the Arabidopsis genome using the RSVSim R package, version 1.7.2 [44], and synthetic reads generated from the modified genome at various levels of sequencing coverage using wgsim [40] (https://github.com/lh3/wgsim). These reads were then used to calculate the true positive, false positive, and false negative TE variant discovery rates for TEPID at various sequencing depths, by running 'tepid-map' and 'tepid-discover' using the simulated reads with the default parameters (Figure 1 - figure supplement 1).

Estimation of sensitivity

Previously published 100 bp paired end sequencing data for L*er* (http://1001genomes.org/data/MPI/MPISchneeberger2011/releases/current/Ler-1/Reads/ ; [22]) was downloaded and analyzed with the TEPID package to identify TE variants. Reads providing evidence for TE variants were then mapped to the *de novo* assembled L*er* genome [21]. To determine whether reads mapped to homologous regions of the L*er* and Col-0 reference genome, the *de novo* assembled L*er* genome sequence between mate pair mapping locations in L*er* were extracted, with repeats masked using RepeatMasker with RepBase-derived libraries and the default parameters (version 4.0.5, http://www.repeatmasker.org). A blastn search was then conducted against the Col-0 genome using the following parameters: '-max-target-seqs 1', '-evalue 1e-6' [23]. Coordinates of the top blast hit for each read location were then compared with the TE variant sites identified using those reads. To estimate false negative rates for TEPID TE absence calls, L*er* TE absence calls were compared with a known set of Col-0-specific TE insertions, absent in L*er* [16]. For TEPID TE insertion calls, we mapped Col-0 DNA sequencing reads [24] to the L*er* PacBio assembly, and identified sites with read evidence reaching the TEPID threshold for a TE insertion call to be made.

Arabidopsis TE variant discovery

We ran the TEPID, including the insertion refinement step, on previously published sequencing data for 216 different Arabidopsis populations (NCBI SRA SRA012474; Schmitz et al. 2013), mapping to the TAIR10 reference genome and using the TAIR9 TE annotation. The '–mask' option was also used to mask the mitochondrial and plastid genomes. We also ran TEPID using previously published transgenerational data for salt stress and control conditions (NCBI SRA SRP045804; [24]), using the '–mask' option to mask mitochondrial and plastid genomes, and the '–strict' option for highly related samples.

TE variant / SNP comparison

SNP information for 216 Arabidopsis accessions was obtained from the 1001 genomes data center (http://1001genomes.org/data/Salk/releases/2013_24_01/; [20]). This was formatted into reference (Col-0 state), alternate, or NA calls for each SNP. Accessions with both TE variant information and SNP data were selected for analysis. Hierarchical clustering of accessions by SNPs as well as

TE variants were used to identify essentially clonal accessions, as these would skew minor allele frequency calculations. A single representative from each cluster of similar accessions was kept, leading to a total of 187 accessions for comparison. For each TE variant with minor allele frequency greater than 3%, the nearest 300 upstream and 300 downstream SNPs with a minor allele frequency greater than 3% were selected. Pairwise genotype correlations (r^2 values) for all complete cases were obtained for SNP-SNP and SNP-TE variant states. r^2 values were then ordered by decreasing rank and a median SNP-SNP rank value was calculated. For each of the 600 ranked surrounding positions, the number of times the TE rank was greater than the SNP-SNP median rank was calculated as a relative LD metric of TE to SNP. TE variants with less than 200 ranks over the SNP-SNP median were classified as low-LD insertions. TE variants with ranks between 200 and 400 were classified as mid-LD, while TE variants with greater than 400 ranks above their respective SNP-SNP median value were classified as variants in high LD with flanking SNPs.

PCR validations

Selection of accessions to be genotyped

To assess the accuracy of TE variant calls in accessions with a range of sequencing depths of coverage, we grouped accessions into quartiles based on sequencing depth of coverage and randomly selected a total of 14 accessions for PCR validations from these quartiles. DNA was extracted for these accessions using Edward's extraction protocol [45], and purified prior to PCR using AMPure beads.

Selection of TE variants for validation and primer design

Ten TE insertion sites and 10 TE absence sites were randomly selected for validation by PCR amplification. Only insertions and absence variants that were variable in at least two of the fourteen accessions selected to be genotyped were considered. For insertion sites, primers were designed to span the predicted TE insertion site. For TE absence sites, two primer sets were designed; one primer set to span the TE, and another primer set with one primer annealing within the TE sequence predicted to be absent, and the other primer annealing in the flanking sequence (Figure 2 - figure supplement 3). Primer sequences were designed that did not anneal to regions of the genome containing previously identified SNPs in any of the 216 accessions [20] or small insertions and deletions, identified using lumpy-sv with the default settings [46](https://github.com/arq5x/lumpy-sv), had an annealing temperature close to 52ºC calculated based on nearest neighbor thermodynamics (MeltingTemp submodule in the SeqUtils python module; [47]), GC content between 40% and 60%, and contained the same base repeated not more than four times in a row. Primers were aligned to the TAIR10 reference genome using bowtie2 [37] with the '-a' flag set to report all alignments, and those with more than 5 mapping locations in the genome were then removed.

PCR

PCR was performed with 10 ng of extracted, purified Arabidopsis DNA using Taq polymerase. PCR products were analysed by agarose gel electrophoresis. Col-0 was used as a positive control, water was added to reactions as a negative control.

mRNA analysis

Processed mRNA data for 144 wild Arabidopsis accessions were downloaded from NCBI GEO GSE43858 [20]. To find differential gene expression dependent on TE presence/absence variation, we first filtered TE variants to include only those where the TE variant was shared by at least 5 accessions with RNA data available. We then grouped accessions based on TE presence/absence variants, and performed a Mann-Whitney U test to determine differences in RNA transcript abundance levels between the groups. We used q-value estimation to correct for multiple testing, using the R qvalue package v2.2.2 with the following parameters: lambda = seq(0, 0.6, 0.05), smooth.df = 4 [48]. Genes were defined as differentially expressed where there was a greater than 2 fold difference in expression between the groups, with a q-value less than 0.01. Gene ontology enrichment analysis was performed using PANTHER (http://pantherdb.org).

DNA methylation data analysis

Processed base-resolution DNA methylation data for wild Arabidopsis accessions were downloaded from NCBI GEO GSE43857 [20], and used to construct MySQL tables in a database.

Rare variant analysis

To assess the effect of rare TE variants on gene expression or DMR DNA methylation levels, we tested for a burden of rare variants in the population extremes, essentially as described previously [28]. For each rare TE variant near a gene or DMR, we ranked the gene expression level or DMR DNA methylation level for all accessions in the population, and tallied the ranks of accessions containing a rare variant. These rank counts were then binned to produce a histogram of the distribution of ranks. We then fit a quadratic model to the counts data, and calculated the R^2 and p-value for the fit of the model.

TE variant and DMR genome-wide association analysis

Accessions were subset to those with both leaf DNA methylation data and TEPID calls. Pairwise correlations were performed for observed data pairs for each TE variant and a filtered set of population C-DMRs, with those C-DMRs removed where more than 15% of the accessions had no coverage. This amounted to a final set of 9,777 C-DMRs. Accession names were then permuted to produce a randomized dataset, and pairwise correlations again calculated. This was repeated 500 times to produce a distribution of expected Pearson correlation coefficients for each pairwise comparison. Correlation values more extreme than any of the 500 permutations were deemed significant.

Data access

TEPID source code can be accessed at https://github.com/ListerLab/TEPID. Code and data needed to reproduce this analysis can be found at https://github.com/timoast/Arabidopsis-TE-variants. L*er* TE variants are available in Figure 1 - source data 1 and 2. TE variants identified among the 216 wild Arabidopsis accessions resequenced by Schmitz et al. (2013) are available in Figure 2 - source data 1, 2 and 3.

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Author contributions

R.L. and T.S. designed the research project. R.L. and J.B. supervised research. T.S. developed and tested TEPID. J.C. performed PCR validations of TE variants. T.S. and S.R.E. performed bioinformatic analysis. Y.K. provided statistical guidance. R.L., T.S., J.B. and S.R.E. prepared the manuscript.

Competing financial interests

The authors declare no competing financial interests.

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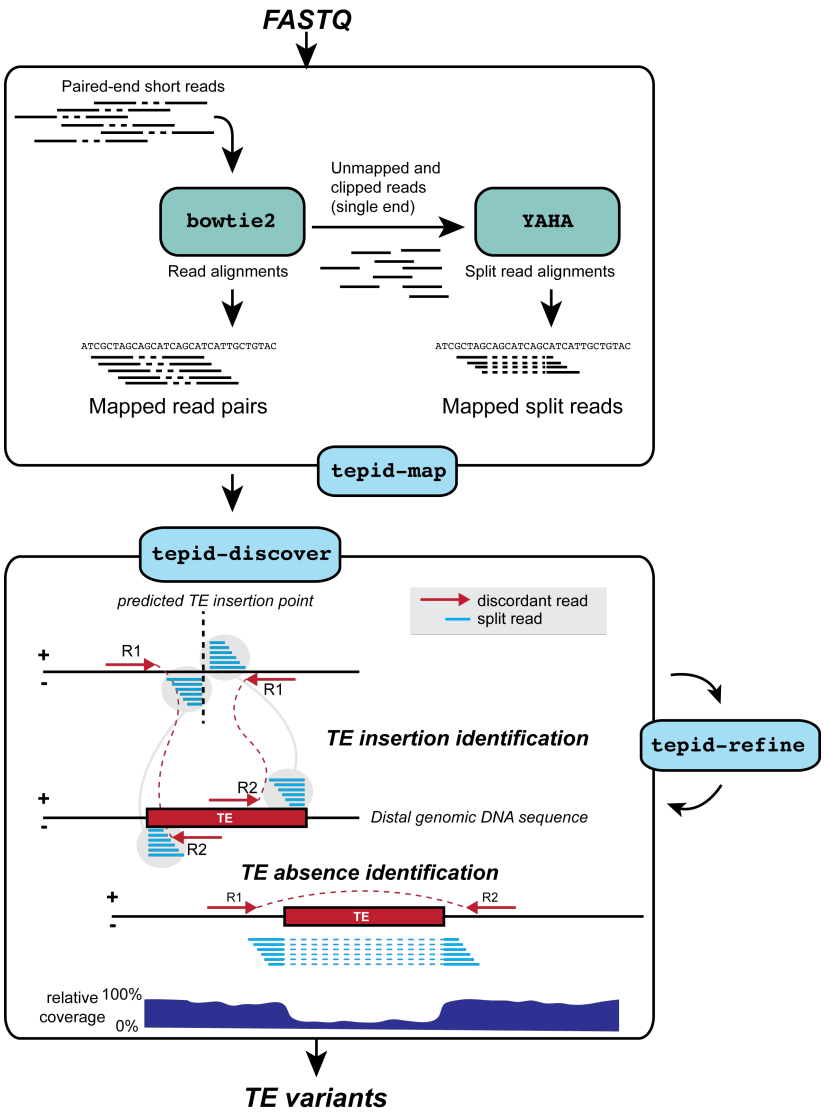


Figure 1: TE variant discovery pipeline

Principle of TE variant discovery using split and discordant read mapping positions. Paired end reads are first mapped to the reference genome using Bowtie2 [37]. Soft-clipped or unmapped reads are then extracted from the alignment and re-mapped using Yaha, a split read mapper [39]. All read alignments are then used by TEPID to discover TE variants relative to the reference genome, in the 'tepid-discover' step. When analyzing groups of related samples, these variants can be further refined using the 'tepid-refine' step, which examines in more detail the genomic regions where there was a TE variant identified in another sample, and calls the same variant for the sample in question using lower read count thresholds as compared to the 'tepid-discover' step, in order to reduce false negative variant calls within a group of related samples.

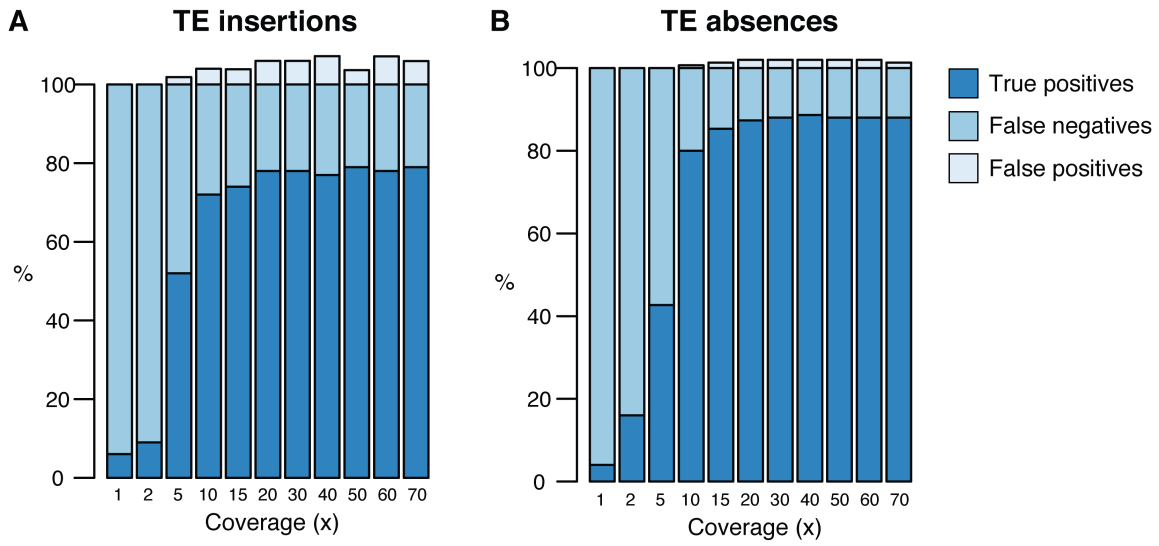


Figure 1: figure supplement 1

Testing of the TEPID pipeline using simulated TE variants in the Arabidopsis Col-0 genome (TAIR10), for a range of sequencing coverage levels. TE insertions (A) and TE absence calls (B).

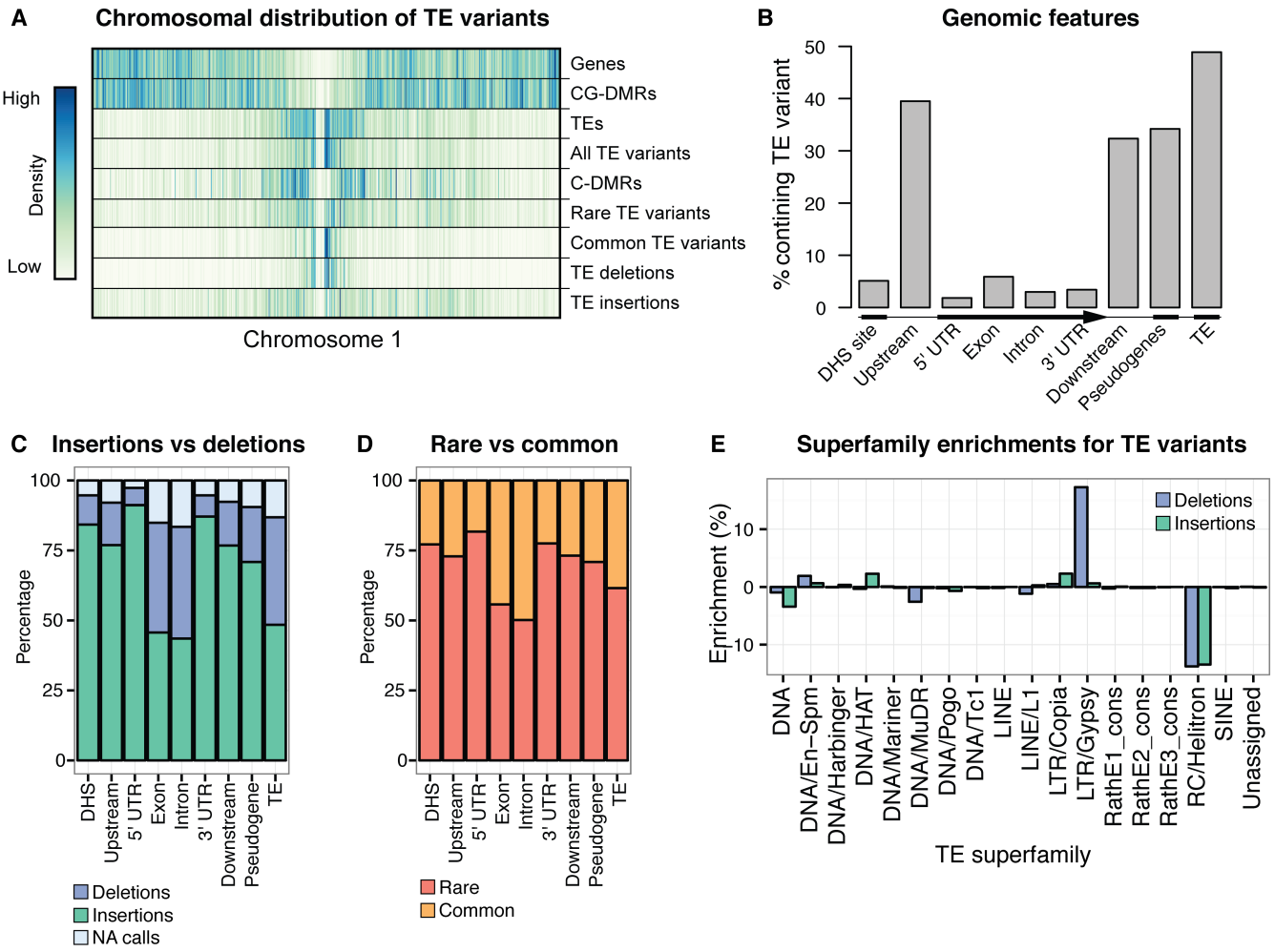


Figure 2: Extensive novel genetic diversity uncovered by TE variant analysis

(A) Distribution of identified TE variants on chromosome 1, with distributions of all Col-0 genes, Col-0 TEs, and population DMRs.

(B) Frequency of TE variants at different genomic features.

(C) Proportion of TE variants within each genomic feature classified as deletions or insertions.

(D) Proportion of TE variants within each genomic feature classified as rare or common.

(E) Enrichment and depletion of TE variants categorized by TE superfamily compared to the expected frequency due to genomic occurrence.

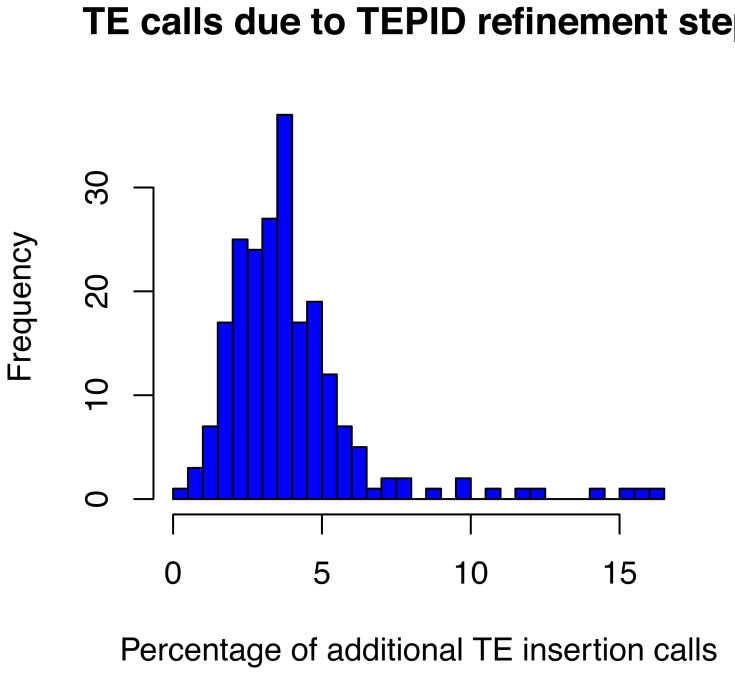


Figure 2: figure supplement 1

Number of additional TE insertion calls made due to the TEPID refinement step for each accession in the population.

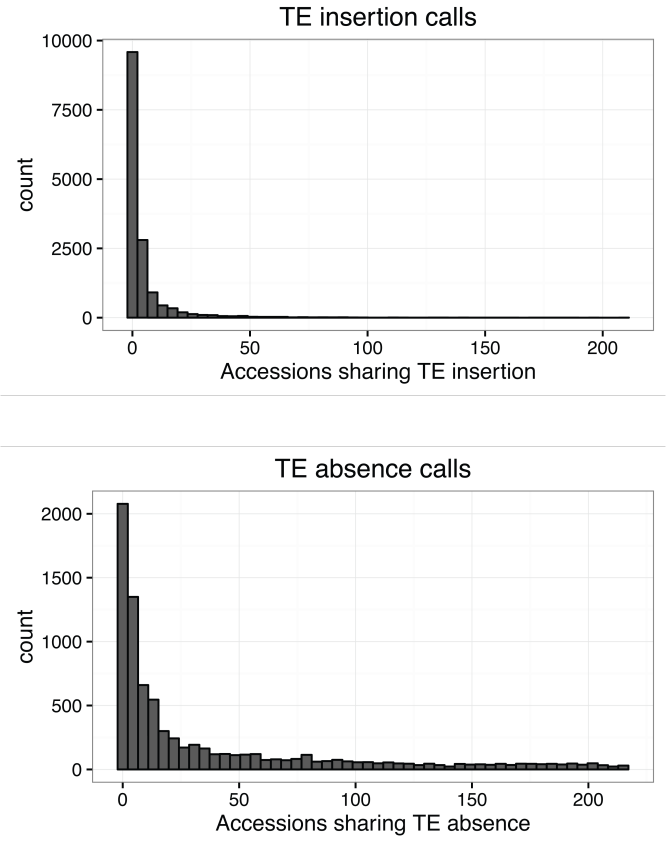


Figure 2: figure supplement 2

748 Minor allele frequency distribution for TE variants.

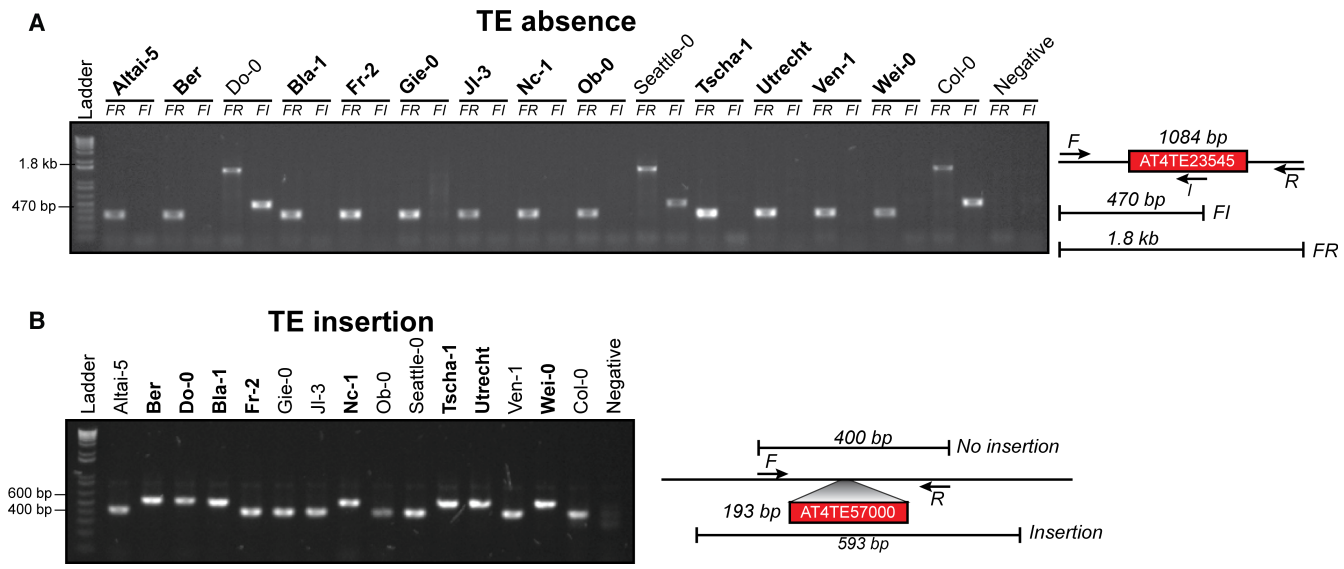


Figure 2: figure supplement 3

(A) PCR validations for a TE absence variant. Accessions that were predicted to contain a TE insertion or TE absence are marked in bold. Two primer sets were used; forward (F) and reverse (R) or internal (I). Accessions with a TE absence will not produce the FI band and produce a shorter FR product, with the change in size matching the size of the deleted TE.

(B) PCR validations for a TE insertion variant. One primer set was used, spanning the TE insertion site. A band shift of approximately 200 bp can be seen, corresponding to the size of the inserted TE.

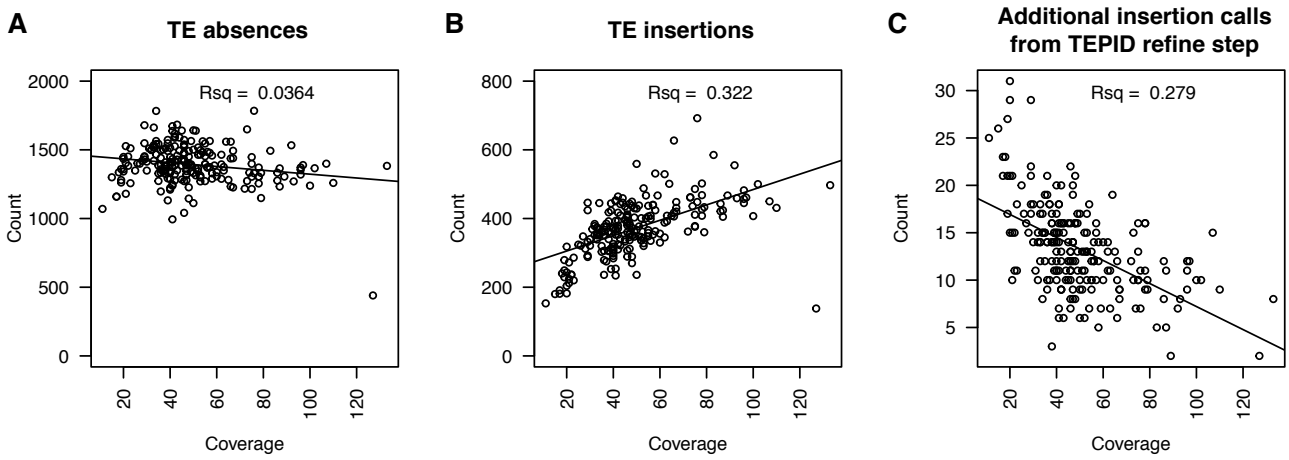


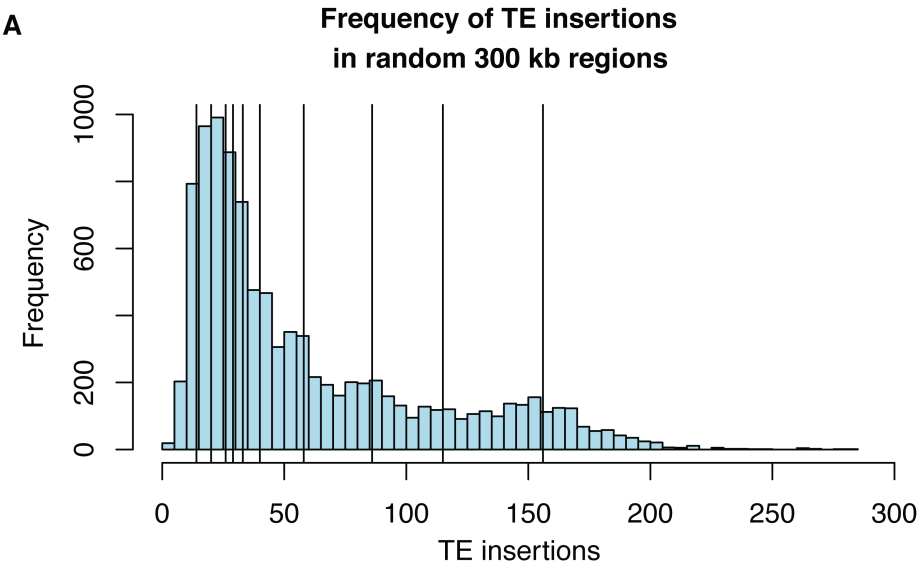
Figure 2: figure supplement 4

(A) Number of TE absence variants identified versus the sequencing depth of coverage for each accession.

(B) Number of TE insertion variants identified versus the sequencing depth of coverage for each accession.

(C) Number of additional TE insertion calls made due to the TEPID refinement step versus sequencing depth of coverage for all accessions.

A



B

chr	start	stop	KEE	TE variants	p-value
chr1	6900000	7200000	kee1	29	0.6304
chr2	4025000	4325000	kee2	156	0.0675
chr3	1800000	2100000	kee3	33	0.5672
chr3	2950000	3250000	kee4	14	0.9172
chr3	16537500	16837500	kee5	115	0.1659
chr3	22375000	22675000	kee6	40	0.4927
chr4	10900000	11200000	kee7	58	0.3589
chr4	15387500	15687500	kee8	26	0.6824
chr5	4612500	4912500	kee9	20	0.802
chr5	10162500	10462500	kee10	86	0.2455

Figure 2: figure supplement 5. Frequency of TE insertion in the *KNOT* region

(A) Number of TE insertion variants within each 300 kb *KNOT ENGAGED ELEMENT (KEE)*, vertical lines) and the number of TE insertion variants found in 10,000 randomly selected 300 kb windows (histogram).

(B) Table showing number of TE insertion variants within each *KEE* region, and the associated p-value determined by resampling 10,000 times.

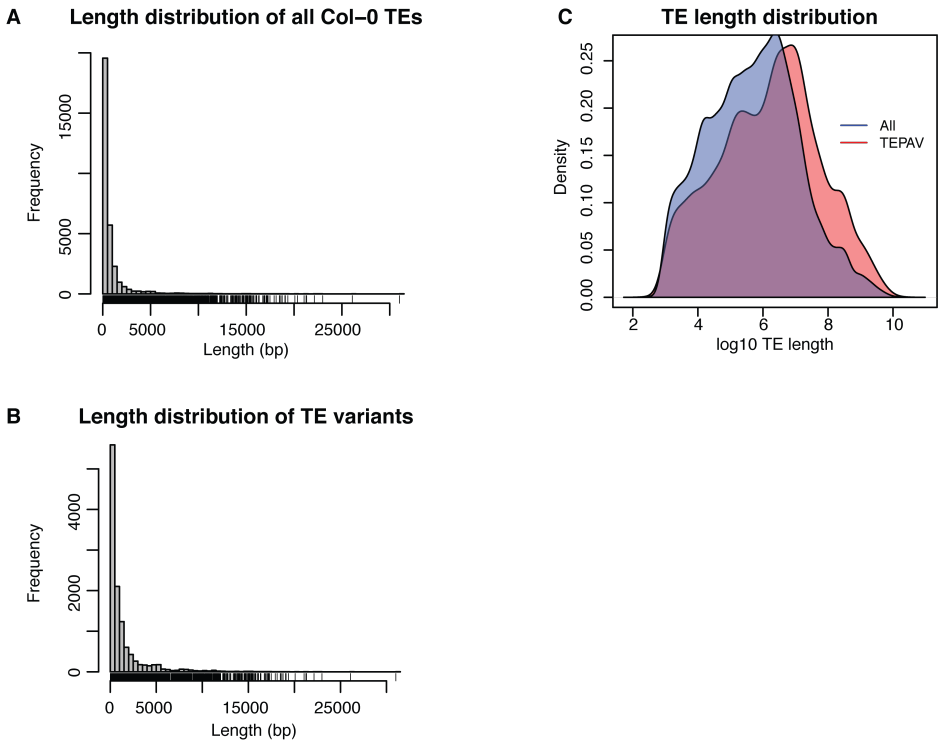


Figure 2: figure supplement 6. Length distribution for all Col-0 TEs and all TE variants

(A) Histogram showing lengths of all annotated TEs in the Col-0 reference genome.

(B) Histogram showing lengths of all TE variants.

(C) Density distribution of log10 TE length for all Col-0 TEs (red) and TE variants (blue).

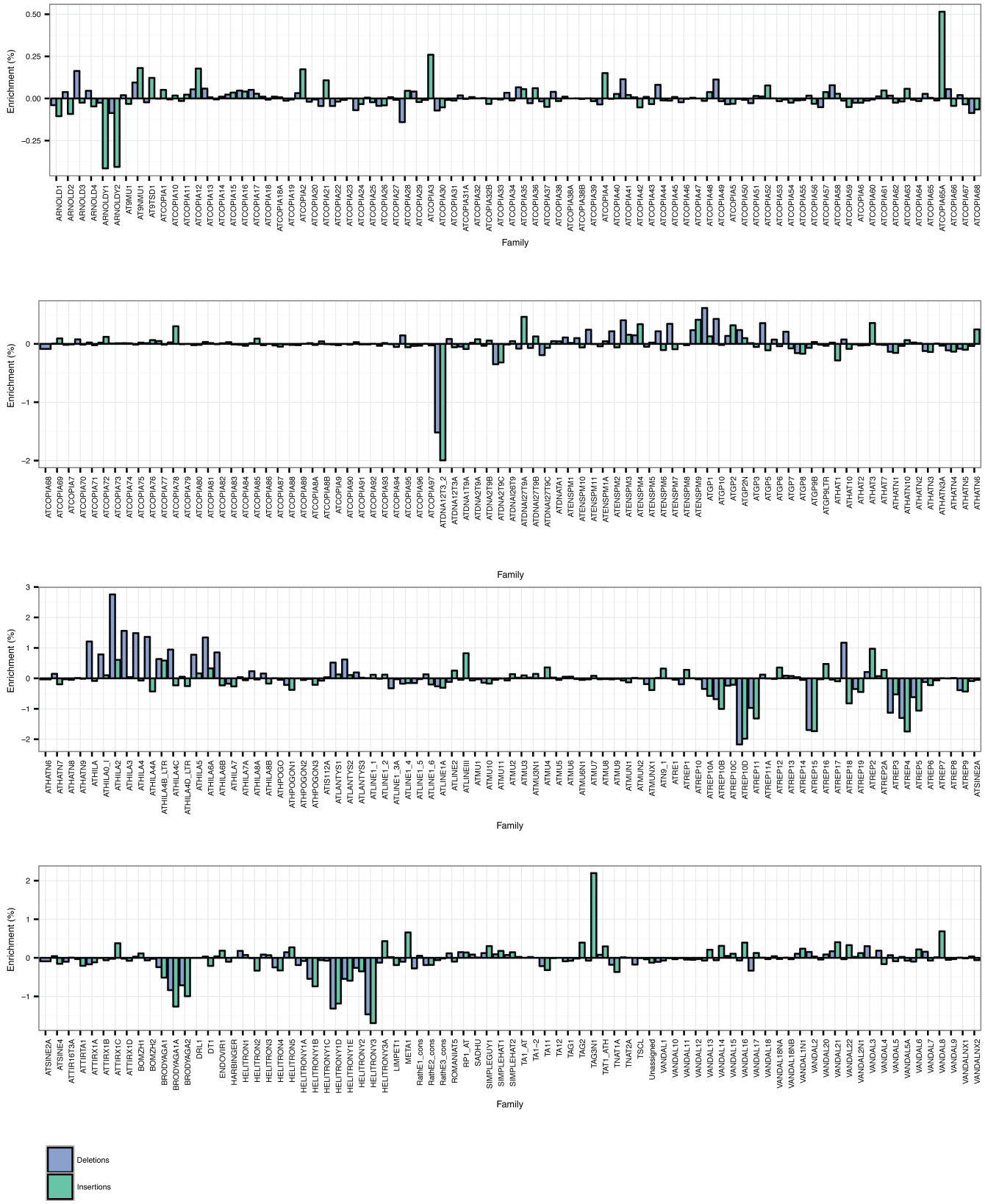


Figure 2: figure supplement 7

TE family enrichments and depletions for TE insertions and TE deletions.

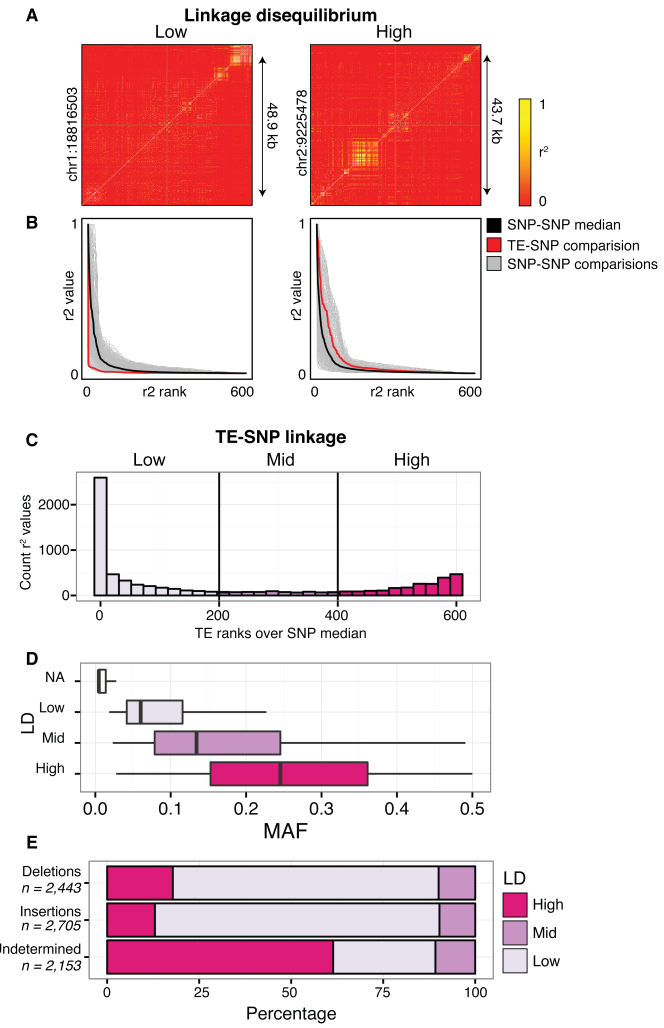


Figure 3: Patterns of TE-SNP linkage

(A) r^2 correlation matrices for individual representative high and low-LD TE variants showing the background level of SNP-SNP linkage.

(B) Rank order plots for individual representative high and low-LD TE variants (matching those shown in A). Red line indicates the median r^2 value for each rank across SNP-based values. Blue line indicates r^2 values for TE-SNP comparisons. Grey lines indicate all individual SNP-SNP comparisons.

(C) Histogram of the number of TE r^2 ranks (0-600) that are above the SNP-based median r^2 value for testable TE variants.

(D) Boxplots showing distribution of minor allele frequencies for each LD category. Boxes represent the interquartile range (IQR) from quartile 1 to quartile 3. Boxplot upper whiskers represent the maximum value, or the upper value of the quartile 3 plus 1.5 times the IQR (whichever is smaller). Boxplot lower whisker represents the minimum value, or the lower value of the quartile 1 minus 1.5 times the IQR (whichever is larger).

(E). Proportion of TE insertions, TE deletions, and unclassified TE variants in each LD category.

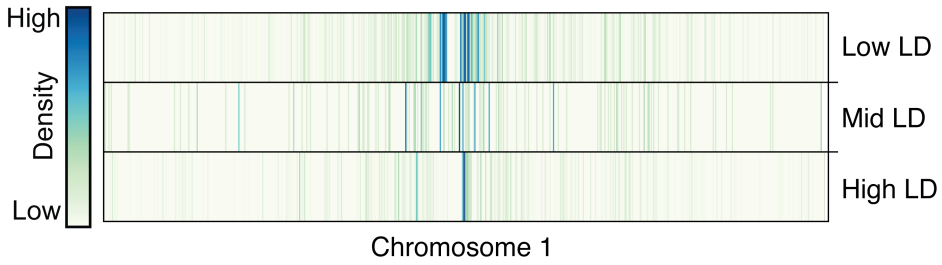


Figure 3: figure supplement 1

Distribution of TE variants across chromosome 1 for each LD category (high, mid, low).

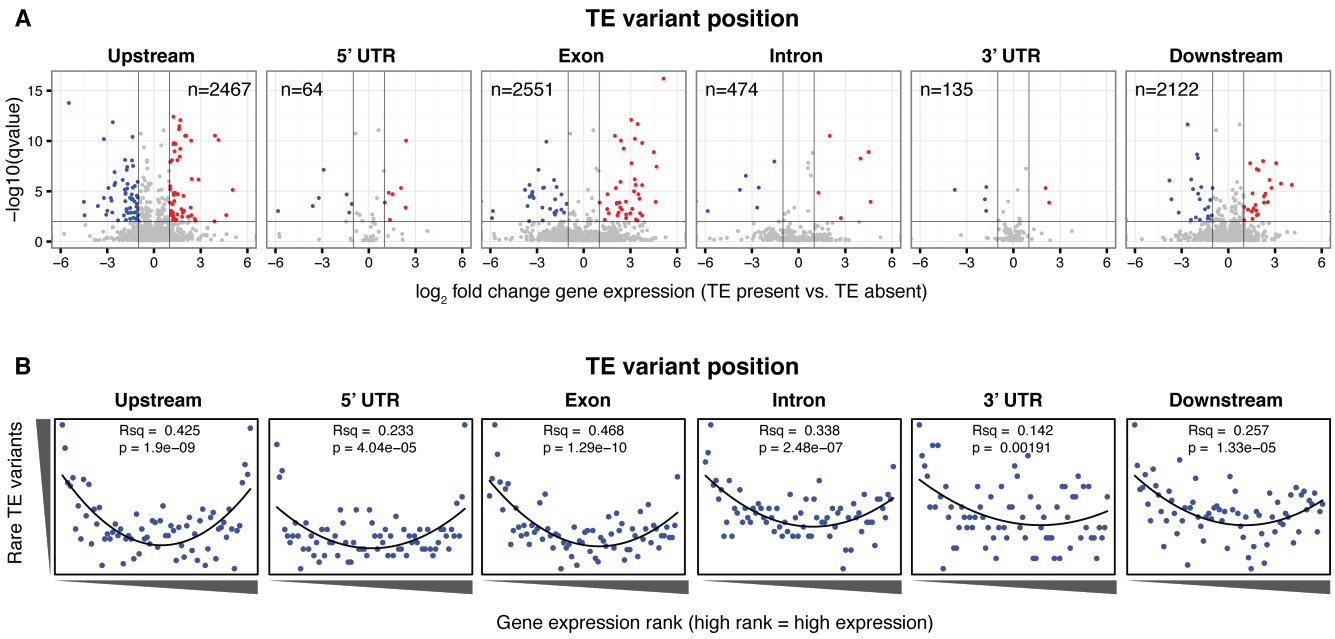


Figure 4: Differential transcript abundance associated with TE variant presence/absence

(A) Volcano plots showing transcript abundance differences for genes associated with TE insertion variants at different positions, indicated in the plot titles. Genes with significantly different transcript abundance in accessions with a TE insertion compared to accessions without a TE insertion are colored blue (lower transcript abundance in accessions containing TE insertion) or red (higher transcript abundance in accessions containing TE insertion). Vertical lines indicate ±2 fold change in FPKM. Horizontal line indicates the 1% FDR.

(B) Relationship between TE rare variant counts and gene expression rank. Plot shows the cumulative number of rare TE variants in equal-sized bins for gene expression ranks, from the lowest-ranked accession (left) to the highest-ranked accession (right). Lines indicate the fit of a quadratic model.

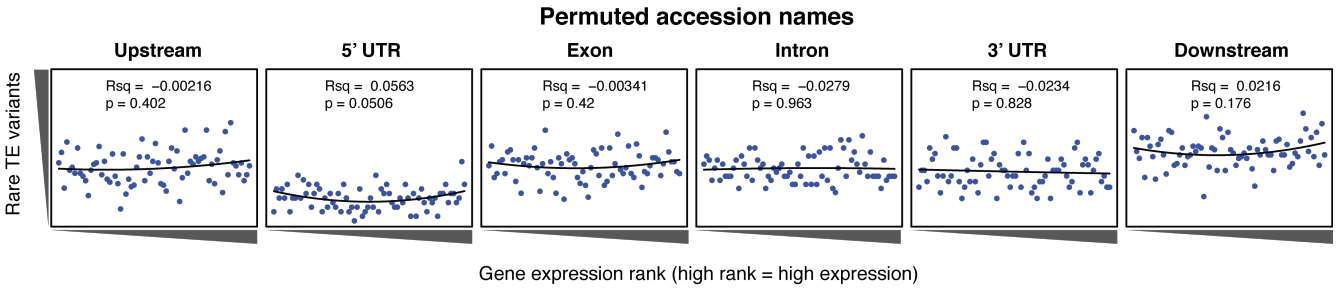


Figure 4: figure supplement 1

Relationship between rare TE variants and gene expression rank as for Figure 4B, for permuted TE variants.

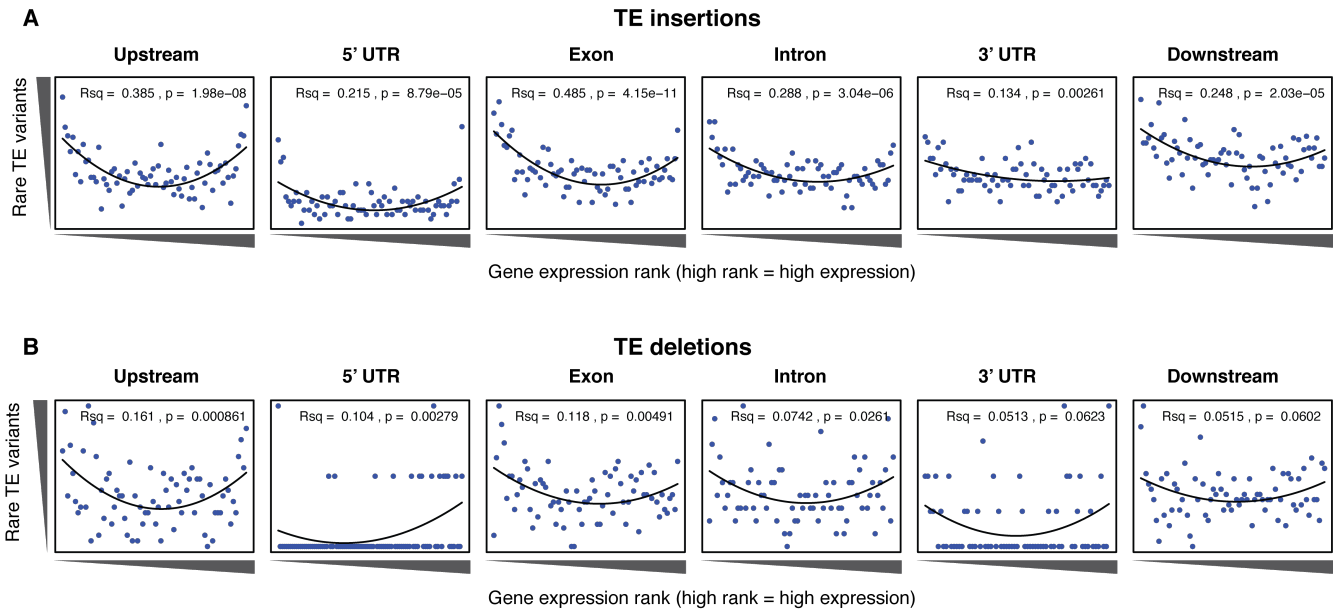


Figure 4: figure supplement 2

Relationship between rare TE variants and gene expression rank as for Figure 4B, for TE insertions (A) and TE deletions (B) separately.

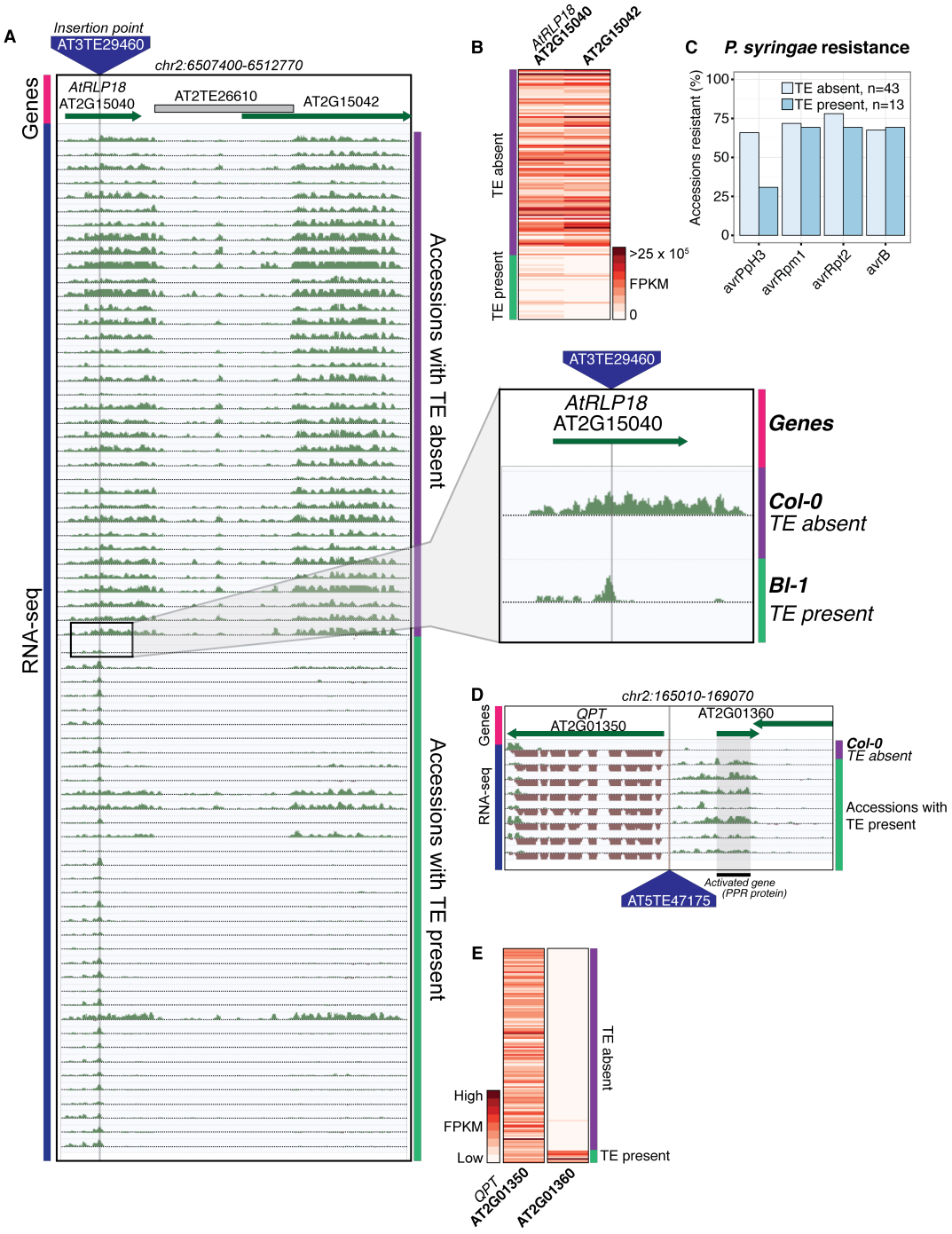


Figure 5: Effects of TE variants on local gene expression

(A) Genome browser representation of RNA-seq data for genes *AtRLP18* (AT2G15040) and a leucine-rich repeat family protein (AT2G15042) for Db-1, containing a TE insertion within the exon of the gene *AtRLP18*, and for a Col-0 (not containing the TE insertion within the exon of *AtRLP18*). Inset shows magnified view of the TE insertion site.

(B) Heatmap showing *AtRLP18* and AT2G15042 RNA-seq FPKM values for all accessions.

(C) Percentage of accessions with resistance to *Pseudomonas syringae* transformed with different *avr* genes, for accessions containing or not containing a TE insertion in* AtRLP18.*

(D) Genome browser representation of RNA-seq data for a PPR protein-encoding gene (AT2G01360) and *QPT* (AT2G01350), showing transcript abundance for these genes in accessions containing a TE insertion variant in the upstream region of these genes.

(E) Heatmap representation of RNA-seq FPKM values for *QPT* and a gene encoding a PPR protein (AT2G01360), for all accessions. Note that scales are different for the two heatmaps, due to the higher transcript abundance of *QPT* compared to AT2G01360. Scale maximum for AT2G01350 is 3.1×10^5, and for AT2G01360 is 5.9×10^4.

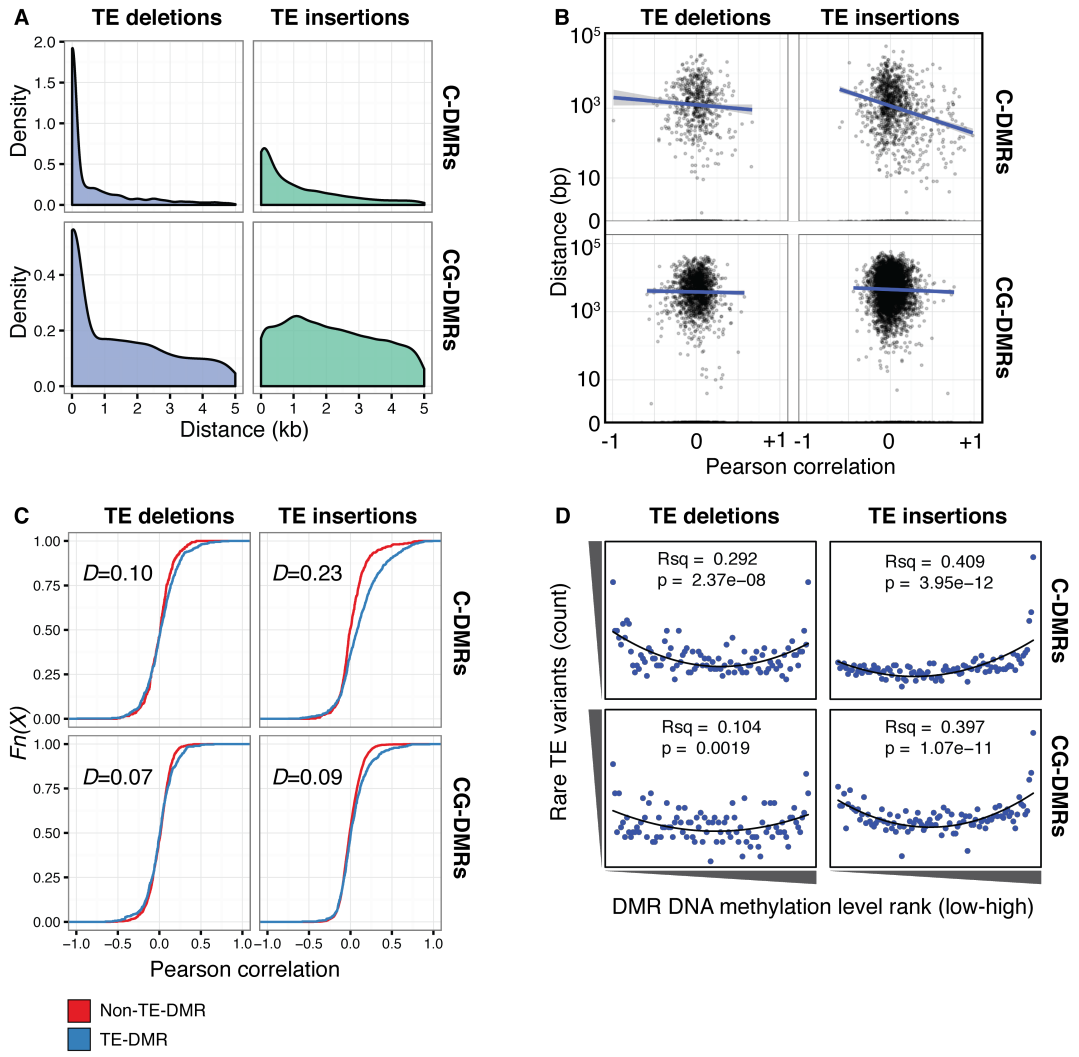


Figure 6: TE variants are associated with nearby DMR methylation levels

(A) Distribution of distances from TE variants to the nearest population DMR, for TE deletions and TE insertion, C-DMRs and CG-DMRs.

(B) Pearson correlation between DMR DNA methylation level and TE presence/absence, for all DMRs and their closest TE variant, versus the distance from the DMR to the TE variant (log scale). Blue lines show a linear regression between the correlation coefficients and the log10 distance to the TE variant.

(C) Empirical cumulative distribution of Pearson correlation coefficients between TE presence/absence and DMR methylation level for TE insertions, TE deletions, C-DMRs and CG-DMRs. The Kolmogorov–Smirnov statistic is shown in each plot, indicated by *D*.

(D) Relationship between rare TE variant counts and nearby DMR DNA methylation level ranks, for TE insertions, deletions, C-DMRs, and CG-DMRs. Plot shows the cumulative number of rare TE variants in equal-sized bins of DMR methylation level ranks, from the lowest ranked accession (left) to the highest ranked accession (right). Lines indicate the fit of a quadratic model, and the corresponding R^2 and p values are shown in each plot.

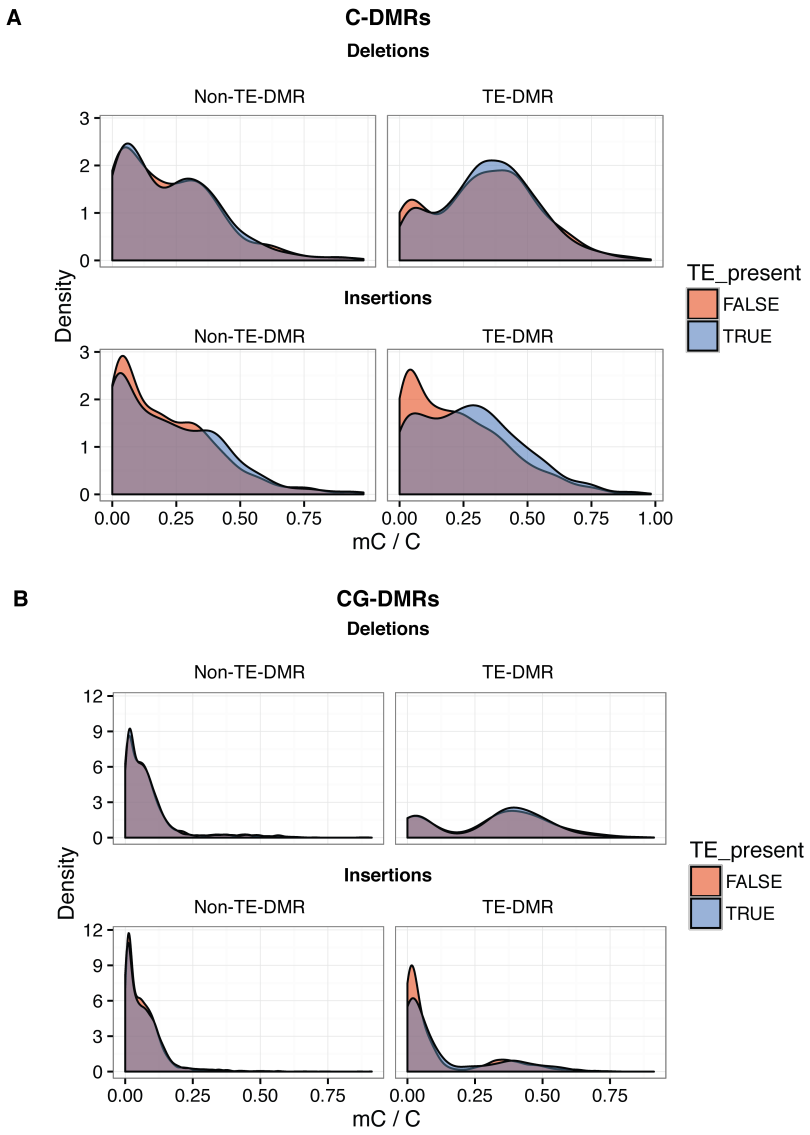


Figure 6: figure supplement 1

828 (A) DNA methylation density distribution at C-DMRs within 1 kb of a TE variant (TE-DMRs) or
829 further than 1 kb from a TE variant (non-TE-DMRs), in the presence or absence of the TE, for
830 TE insertions and TE deletions.

831 (B) As for A, for CG-DMRs.

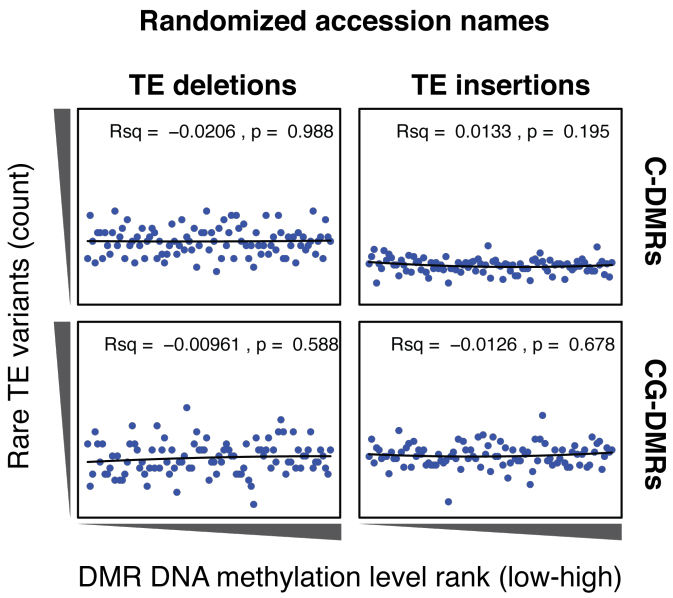


Figure 6: figure supplement 2

Cumulative number DMR methylation level ranks for DMRs near rare TE variants with accessions selected at random. Lines indicate the fit of a quadratic model, and the corresponding R^2 and p values are shown in each plot.

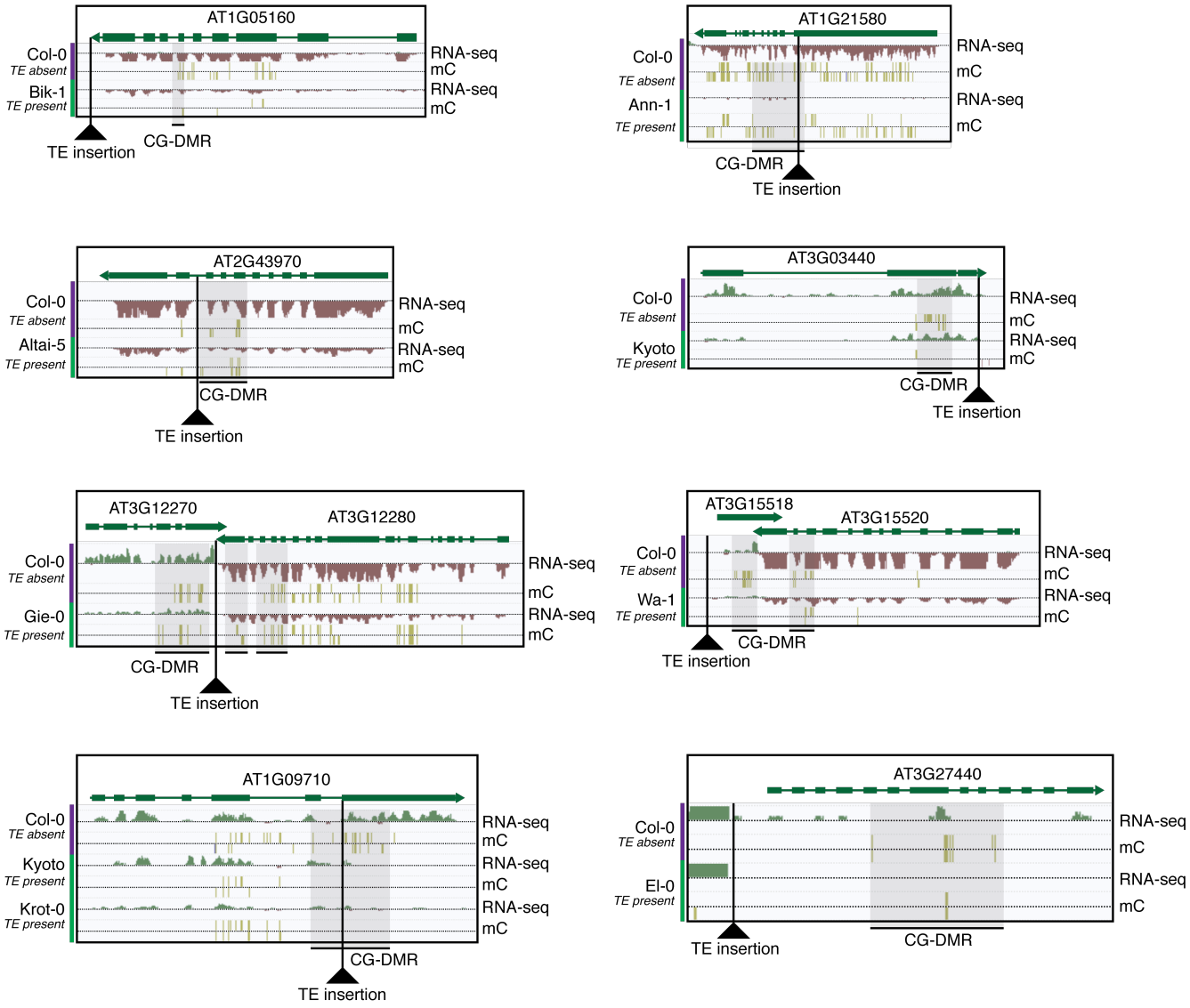


Figure 6: figure supplement 3

Selected examples of TE insertions apparently associated with transcriptional downregulation of nearby genes and loss of gene body CG methylation leading to the formation of a CG-DMR.

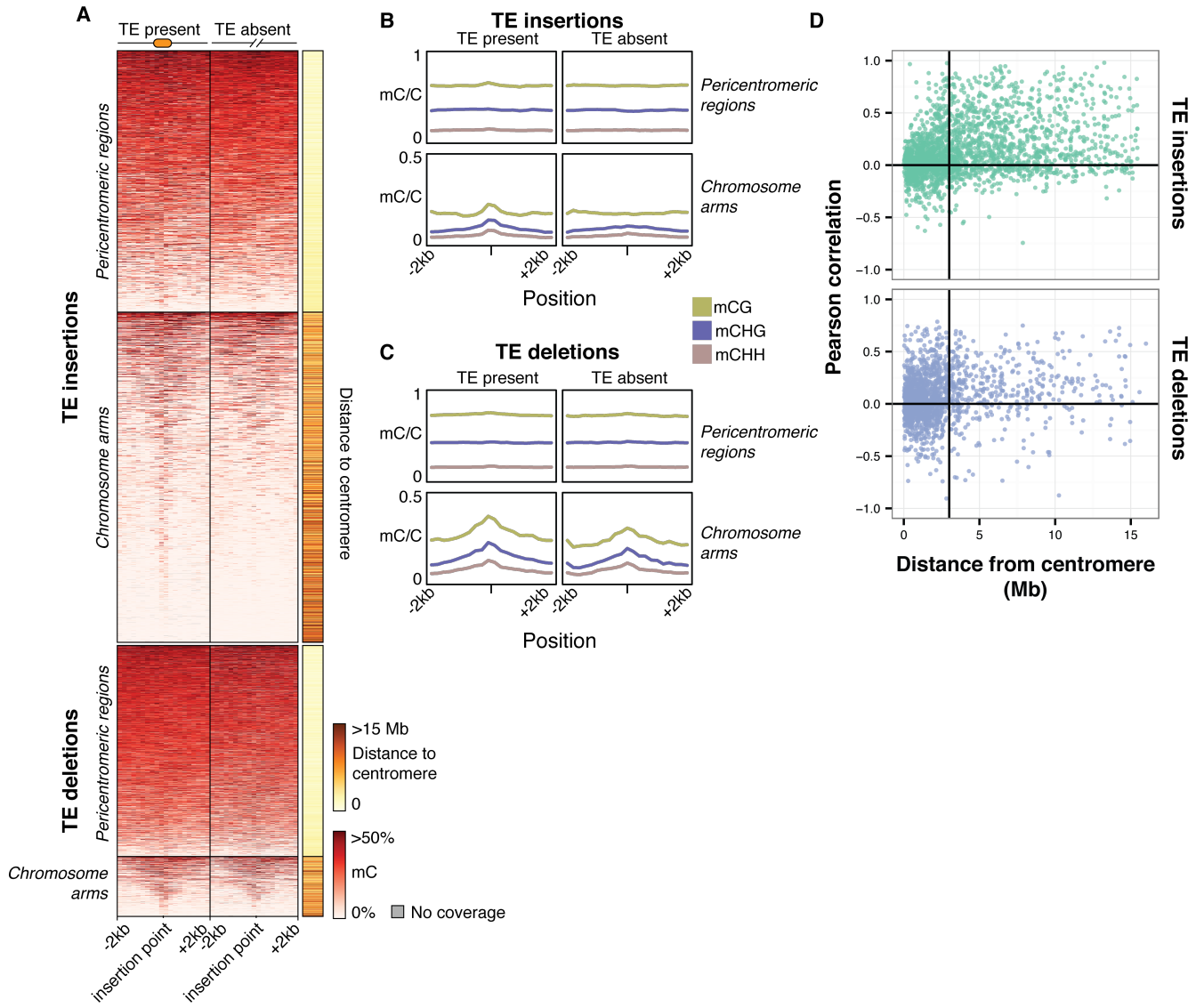


Figure 7: Local patterns of DNA methylation surrounding TE variant sites

(A) Heatmap showing DNA methylation levels in 200 bp bins flanking TE variant sites, +/- 2 kb from the TE insertion point. TE variants were grouped into pericentromeric variants (<3 Mb from a centromere) or variants in the chromosome arms (>3 Mb from a centromere).

(B) Line plot showing the DNA methylation level in each sequence context for TE insertion sites, +/- 2 kb from the TE insertion point.

(C) As for B, for TE deletions.

(D) Distribution of Pearson correlation coefficients between TE presence/absence and DNA methylation levels in the 200 bp regions flanking TE variant, ordered by distance to the centromere.

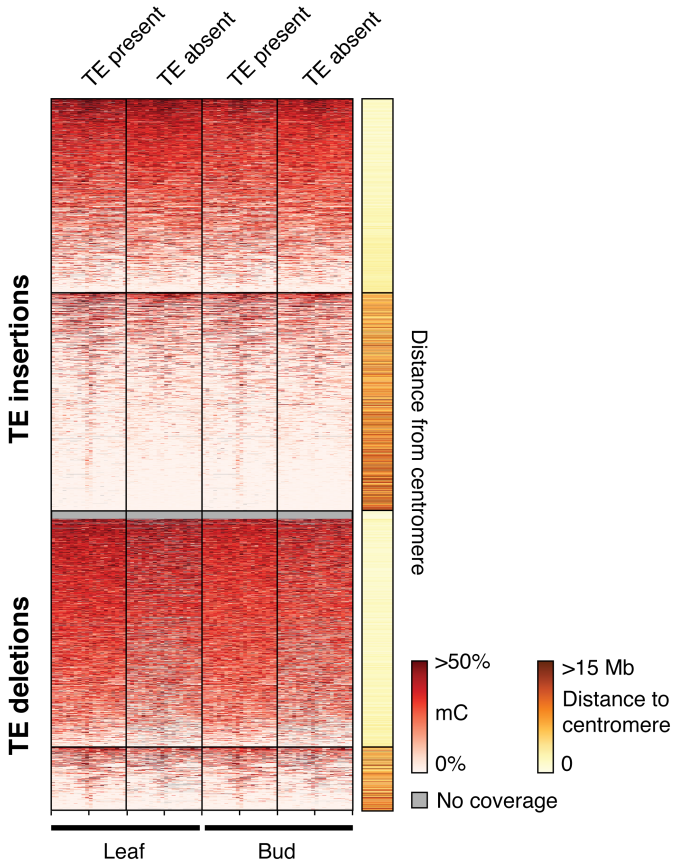


Figure 7: figure supplement 1

Heatmap showing DNA methylation levels in 200 bp bins flanking TE variant sites in the 12 accessions with DNA methylation data for both leaf and bud tissue, +/- 2 kb from the TE insertion point. TE variants were grouped into pericentromeric variants (<3 Mb from a centromere) or variants in the chromosome arms (>3 Mb from a centromere).

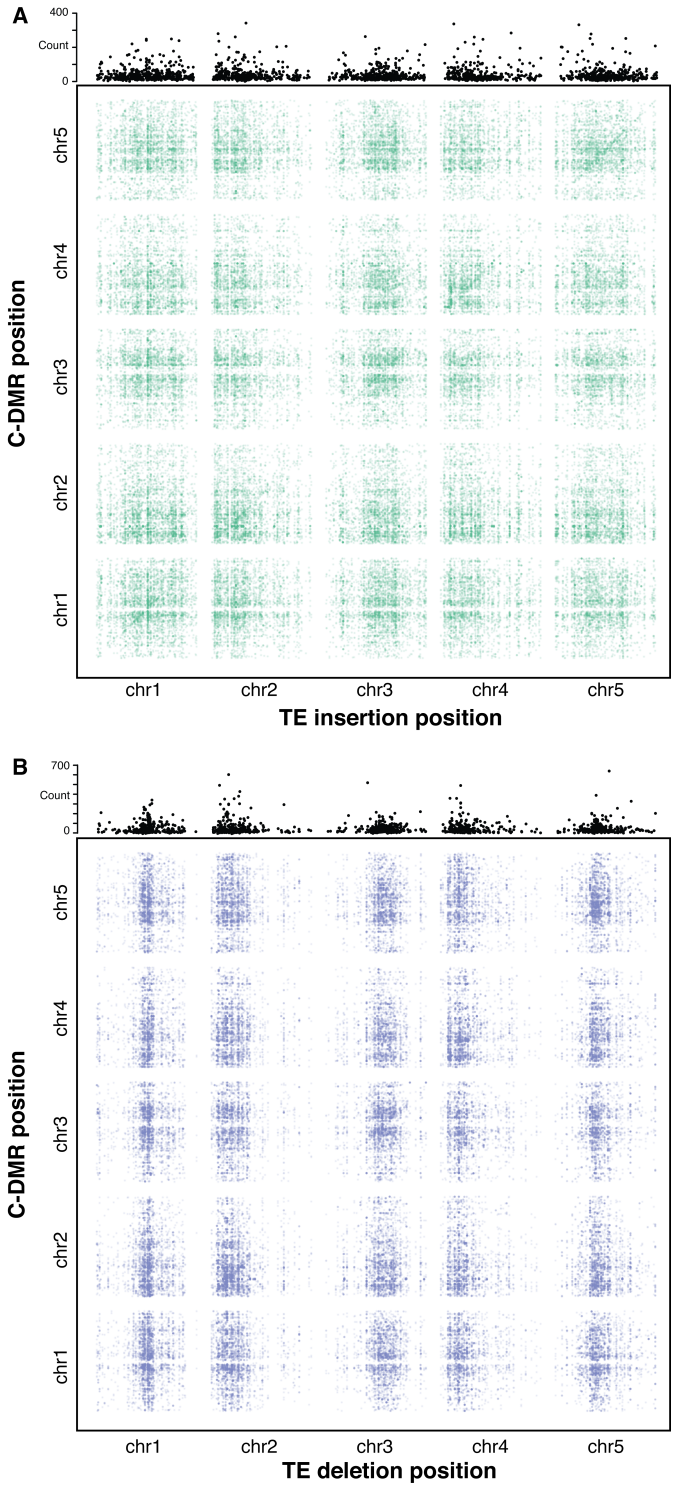


Figure 8: Association scan between TE variants can C-DMR methylation variation

(A) Significant correlations between TE insertions and C-DMR DNA methylation level. Points show correlations between individual TE-DMR pairs that were more extreme than any of 500 permutations of the DMR data. Top plots show the total number of significant correlations for each TE insertion across the whole genome.

(B) As for (A), for TE deletions.

Table 1: Mapping of paired-end reads providing evidence for TE presence/absence variants in the L*er* reference genome

	Concordant	Discordant	Split	Unmapped	Total
Col-0 mapped	0	993	9513	0	10206
L*er* mapped	10073	92	34	7	10206

Note: Discordant and split read categories are not mutually exclusive, as some discordant reads may have one read in the mate pair split-mapped.

Table 2: Summary of TE variant classifications

TEPID call	TE classification	Count
Insertion	NA	310
	Insertion	14689
	Deletion	8
Absence	NA	1852
	Insertion	388
	Deletion	5848

Table 3: Percentage of DMRs within 1 kb of a TE variant

	C-DMRs			CG-DMRs		
	Observed	Expected	95% CI	Observed	Expected	95% CI
TE deletions	8.7	16	0.0078	4.3	16	0.0041
TE insertions	36	26	0.0089	9.4	26	0.0047
NA calls	3.4	6.2	0.0052	1.7	6.2	0.0027
Total	48	41	0.01	15	42	0.0054