

Phased T2T genome of a tetraploid grapevine reveals segmental allopolyploid origin and allele-specific alternative splicing during fruit development

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Abstract

The tetraploid table grape Kyoho, a cultivar of major global economic importance, has long presented a challenge in understanding its genetic architecture and the regulatory landscapes between its different ancestral genomes. Here, we assemble a haplotype-resolved, near-complete reference genome (480.32-489.47 Mb, contig N50 17.5-23.3 Mb, with a total of only 11 gaps) to investigate regulatory dynamics through integrative multiomic analyses during fruit development. Comparative evolutionary genomics classifies Kyoho as a segmental allotetraploid with a complex mosaic genome, composed of ~71% *V. vinifera* (Vv) and ~29% *V. labrusca* (Vl) ancestry. A total of 67.3% of genes are shared among the four haplotypic genomes with 9.8% haplotype specific genes. Transcriptomic analysis revealed that among 17,750 pairs of ancestral alleles exhibiting directional bias, only 4,086 (23.4%) maintained a consistent ancestral dominance direction across all tissues. Similarly, comparisons of DNA methylation profiles showed no significant differences between genome ancestries Vl (0.009~0.676): Vv (0.010~0.670). However, alternative splicing (AS) analyses demonstrated that allele-specific splicing changes were correlated with ancestral origin across tissues (Vl: 1,529 vs. Vv: 2,052). Integrative omics identified 2,307 differentially expressed ancestral alleles that showed no significant genomic variants or methylation bias but exhibited AS bias. These genes include key regulators of fruit ripening, including *ZEP*, *PIN*, *DOG1* and *MYB*. Overall, this work delivers a landmark genomic resource and characterizes its regulatory architecture, facilitating functional genomic studies and guiding precision polyploid breeding in grapes.

Keywords: Polyploid, Structural variation, Allelic genes, Multi-omics, Alternative splicing, Grape breeding

Introduction

Polyploidy is an important force in plant evolution and crop breeding (Cheng *et al.*, 2025). Polyploids are typically classified into two distinct types: autopolyploid and allopolyploid. Autopolyploids arise from whole-genome duplication within a single species, exhibiting multivalent chromosome pairing and tetrasomic inheritance (Xue *et al.*, 2022, Sun *et al.*, 2022).

1 Conversely, allopolyploids originate from hybridization between divergent species. They exhibit
2 strictly bivalent pairing and tetrasomic inheritance (Zhuang *et al.*, 2019, Yoo *et al.*, 2013, Pfeifer
3 *et al.*, 2014). However, rather than a strict dichotomy, polyploidy actually exists along a continuous
4 spectrum. This spectrum is often bridged by intermediate forms such as segmental allopolyploids,
5 which arise from complex hybridization and display mixed meiotic behaviors (Bertioli *et al.*, 2019,
6 Wu *et al.*, 2025, Song *et al.*, 2023). Polyploid crops often exhibit enhanced vigor and stress
7 tolerance but pose significant challenges in biological studies and breeding applications (Wang *et al.*,
8 2024a, Zhuang *et al.*, 2019, Yoo *et al.*, 2013, Pfeifer *et al.*, 2014, Xue *et al.*, 2022). The
9 coexistence of highly similar homologous sequences of different ancestral origins and frequent
10 structural variations complicates the analysis of regulatory networks, obscuring the link between
11 genotype and phenotype in polyploids (Wang *et al.*, 2024a, Zhuang *et al.*, 2019, Yoo *et al.*, 2013,
12 Pfeifer *et al.*, 2014, Xue *et al.*, 2022). Recent breakthroughs in phased genomics have underscored
13 that key agronomic traits are often governed by specific haplotype interaction (Brinton *et al.*, 2020,
14 Todesco *et al.*, 2020, Walkowiak *et al.*, 2020). Thus, dissecting polyploid genomes at the haplotype
15 level is paramount for deciphering the genetic basis of complex traits.

16 The advent of phased genomics is revealing that the boundary between auto- and allopolyploidy
17 is far more blurred than previously thought (Zhang *et al.*, 2025, Song *et al.*, 2024b, Sun *et al.*,
18 2022). Several economically important crops are not simple combinations of distinct subgenomes
19 but possess chimeric genomic architectures (Zhang *et al.*, 2025, Bertioli *et al.*, 2019, Wu *et al.*,
20 2025, Song *et al.*, 2023). In such segmental allopolyploids, a single homologous chromosome
21 group may act as a mosaic, containing haplotypes of mixed ancestry alongside those from a single
22 progenitor (Zhang *et al.*, 2025, Bertioli *et al.*, 2019, Wu *et al.*, 2025, Song *et al.*, 2023). This
23 structural mosaicism creates a unique regulatory landscape where divergent ancestral regulatory
24 codes must coexist (Xue *et al.*, 2022, Sun *et al.*, 2022, Zhang *et al.*, 2022, Zhang *et al.*, 2019). How
25 do epigenetic mechanisms, such as DNA methylation, and post-transcriptional processes, like
26 alternative splicing, coordinate across these distinct ancestral haplotypes? The hybridization
27 triggers epigenetic reprogramming (Xue *et al.*, 2022, Sun *et al.*, 2022) and that splicing expands
28 transcriptomic diversity (Zhang *et al.*, 2022, Zhang *et al.*, 2019), the rules governing how these
29 regulatory layers are partitioned and synchronized within a single, structurally chimeric genome
30 remain largely unexplored.

Grapevine (*Vitis vinifera* L.) is a crop of immense economic and cultural importance, yet genomic research has predominantly focused on diploid samples (Shi *et al.*, 2023, Wang *et al.*, 2024b, Peng *et al.*, 2025, Liu *et al.*, 2024, Xiao *et al.*, 2025), leaving superior polyploid cultivars underexplored. The Kyoho grape, one of the world's most cultivated variety (OIV, 2017), epitomizes the complex polyploidy described above. As a tetraploid derived from an interspecific cross between *V. vinifera* and *V. labrusca*—species lacking strong reproductive barriers. Strikingly, our genomic analysis reveals that this hybrid origin has shaped Kyoho into a classic segmental allopolyploid. Rather than exhibiting uniform genetic behavior, it displays a complex mixture of both tetrasomic and disomic inheritance. This mixed meiotic behavior drives extensive but localized recombination, creating a deeply blurred genomic boundary and a highly mosaic chromosomal architecture. Consequently, it serves as an ideal system for dissecting the interplay between mosaic ancestry and regulatory evolution. Although its hybrid origin is known, the precise architecture of its genome—specifically, how its constituent haplotypes trace back to their ancestors and how these distinct ancestries navigate the regulatory landscape—remains an open question.

Here, we present a haplotype-resolved, near-complete genome assembly of Kyoho to dissect its genomic origin and regulatory landscape of a complex hybrid tetraploid. By integrating genomics, epigenomics, and transcriptomics (RNA-seq and Iso-Seq), we address three fundamental questions. (1) What is the precise genomic architecture of this complex tetraploid, and does it exhibit the mosaic ancestry characteristic of segmental allopolyploids? (2) How are distinct regulatory strategies, including gene expression, DNA methylation and alternative splicing, partitioned between haplotypes of *V. vinifera* versus *V. labrusca* ancestry? (3) Is there a mechanistic link between allele-specific DNA methylation and the regulation of alternative splicing across these ancestral alleles, driving their functional specialization? This work provides a new paradigm for understanding the interplay between mosaic genomic ancestry and regulatory evolution in polyploid crops.

Results

A haplotype-resolved near-complete genome assembly for tetraploid Kyoho table grape

To achieve a high-quality, phased genome assembly for the tetraploid table grape Kyoho, we generated 163.9 Gb of PacBio HiFi (~72×), 42.49 Gb of Oxford Nanopore (ONT) (~34×), and 200 Gb of Hi-C (~110×) data. K-mer analysis (21-mers) estimated a haploid genome size of

approximately 435.36 Mb with a high heterozygosity rate of 3.6% (Fig. 1C, Supplementary Fig. 1B). Fluorescence in situ hybridization (FISH) analysis confirmed the tetraploid state with 76 chromosomes (Fig. 1E). By integrating these datasets, we constructed a haplotype-resolved, near-complete reference genome spanning 1.81 Gb, anchored into 76 pseudochromosomes (4 sets of 19). The assembly achieved exceptional continuity, with only 11 gaps remaining across the entire tetraploid genome. The four phased assemblies (designated hap1-hap4) exhibited comparable sizes (480.32-489.47 Mb) and high contiguity, with contig N50 values ranging from 17.5 Mb to 23.3 Mb (Table S1). Importantly, the labels “hap1” through “hap4” are utilized strictly as a technical convention to distinguish the four phased copies of each chromosome within our assembly; they are not intended to imply the existence of four biologically distinct, genome-wide co-segregating haplotypes.

We rigorously validated the quality of genome assembly, confirming its near-telomere-to-telomere (T2T) status. First, the size of each haplotype closely matched the *V. vinifera* reference (PNT2T, ~494.87 Mb), and sequencing reads from all platforms (ONT, HiFi, Illumina) mapped back with rates approaching 100% (Supplementary Fig. S1A). Second, the Hi-C contact maps displayed clear diagonal lines with negligible off-diagonal noise, confirming the precise ordering and orientation of contigs within chromosomes (Fig. 1F). Third, the high completeness of the assembly was demonstrated by Benchmarking Universal Single-Copy Orthologs (BUSCO) scores averaging 97.95% for all haplotypes (range: 97.60%-98.30%) (Fig. 1D, Supplementary Fig. S1C) and high LTR Assembly Index (LAI) scores (range: 16.42-19.55). Most notably, the structural integrity of the assembly allowed for the precise delineation of centromeres for all 76 chromosomes and the end-to-end reconstruction of 122 out of 152 expected telomeres (Supplementary Table S1), providing a robust foundation for downstream analyses.

With a high-quality backbone established, we characterized the genomic landscape of each haplotype. Repeat sequences constituted approximately two-thirds of the genome (range: 66.32%-67.14%), with Long Terminal Repeats (LTRs) being the dominant class (range: 31.70%-32.49%) (Supplementary Table S2). We predicted an average of 29,742 high-confidence protein-coding genes per haplotype (range: 29,599-29,960) (Supplementary Table S1 and Table S3). This comprehensive annotation provides the necessary resolution to dissect the interplay between genomic elements and regulatory layers discussed below.

1 To unravel the genome-wide diversity of Kyoho, we aligned the four haplotypes to the PNT2T
2 reference, identifying 14,079,340 SNPs, 2,917,719 Indels, and 373,745 structural variants (SVs
3 >50 bp), with SNPs and Indels affecting ~4% of the genome. Private variation loads were unevenly
4 distributed: we detected 2,278,773, 2,430,253, 1,071,101, and 997,718 haplotype-specific SNPs
5 in hap1-hap4, respectively, along with 547,559, 557,280, 261,110, and 244,825 specific Indels; and
6 70,395, 72,548, 40,854, and 40,378 specific SVs (Supplementary Fig. S2, Fig. S3A, Table S4 and
7 Table S5). PCA confirmed significant divergence from the reference, while chromosomal
8 clustering analyses revealed distinct three-group patterns for SNPs (Chr01, 10, 13, 18), SVs
9 (Chr01, 04, 10, 13, 19), and Indels (Chr01, 10, 11, 13, 18, 19). In total, 59,747 genes harbored
10 haplotype-specific variants, with 34% containing multiple variant types (Supplementary Fig. S3B
11 and Table S6).

12 To further assess genome-wide diversity, pairwise whole-genome alignments revealed a
13 chromosome-stratified affinity map. While Chr04, 14, and 17 showed high global conservation,
14 other chromosomes exhibited mosaic haplotype grouping: hap1 was distinct on Chr13 and 15;
15 hap2 was divergent on Chr01; and hap2/hap3 clustered closely on Chr02, 06, 08, 10, 16, and 18,
16 where hap1/hap4 also showed high consistency on Chr18. This heterogeneous pattern, reflecting
17 complex ancestry and selection, precludes simple autopolyploid or allopolyploid classification.
18 Finally, although we detected 67.3% of genes retained in four copies and 9.8% of genes retaining
19 as single copies (Supplementary Fig. S3C and Table S7), 8,765 single-copy genes were overlapped
20 with private variants (Supplementary Fig. S3D), indicating that significant haplotype-specific
21 loads persist even at loci under strong functional constraints.

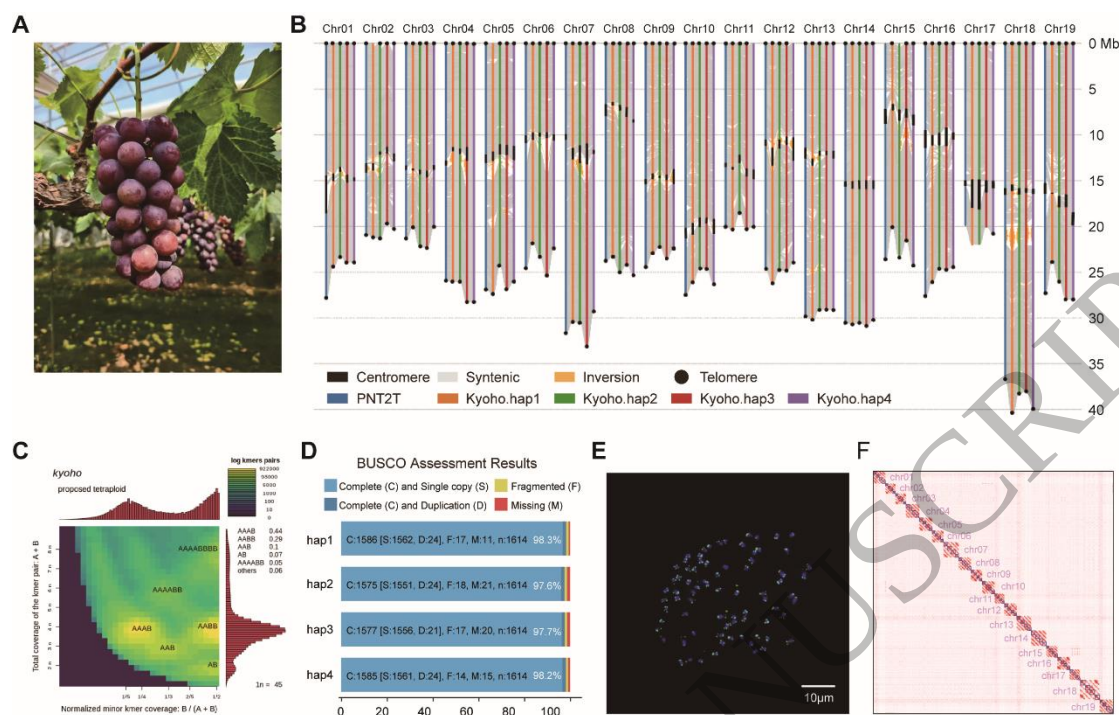


Figure 1. Genome assembly and characterization of the tetraploid Kyoho grape. A. Photograph of a mature Kyoho grape cluster. B. Chromosome-level collinearity analysis between the four Kyoho haplotypes and the reference genome (PNT2T). C. Smudgeplot analysis based on PacBio HiFi reads. D. BUSCO assessment of genome completeness for the four haplotypes. E. Fluorescence in situ hybridization (FISH) image of Kyoho chromosomes. The image shows the Kyoho chromosome set, with chromosomes counterstained in blue with DAPI. F. Hi-C interaction heatmap of the Kyoho genome.

The evolutionary genomic origin of Kyoho

To resolve the evolutionary origin of the Kyoho tetraploid, we first constructed phylogenomic trees for each chromosome using 31 *Vitis* taxa. While the genome overall reflects its known hybrid parentage from Ishihara Wase (*V. labrusca*) and Centennial (*V. vinifera*) (OIV, 2017) (Fig. 2A, Supplementary Fig. S4), a closer examination revealed striking heterogeneity in ancestral inheritance. While some chromosomes (such as Chr04) showed all four haplotypes grouping with the *V. vinifera* reference (PNT2T), indicating a purely *V. vinifera* origin for this homologous group. In contrast, other chromosomes displayed distinct mixed ancestries. For instance, the phylogenetic tree for Chr01 revealed a clear segregation: while haplotypes hap1, hap3, and hap4 clustered with *V. vinifera* (Group IV), hap2 diverged significantly, grouping with *V. labrusca* (Group I) (Fig. 2A).

1 These conflicting phylogenetic signals across chromosomes (Fig. 2A, Supplementary Fig. S4)
 2 provided the initial evidence that Kyoho is not a uniform hybrid but possesses a complex, mosaic
 3 genomic structure.

4 By integrating these phylogenetic patterns with local genetic distance calculations, we deciphered
 5 the precise genomic architecture of Kyoho. Our analysis confirms that the genome is a complex
 6 mosaic lacking the uniform subgenome separation typical of strict allopolyploids. This mosaicism
 7 operates at two distinct organizational scales. First, at the chromosomal level, we observed varied
 8 combinations of parental haplotypes, ranging from pure *V. vinifera* groups to mixed configurations
 9 (Fig. 2A, Supplementary Fig. S4). Second, and most critically, we uncovered extensive intra-
 10 chromosomal recombination. At least 9 of the 19 chromosome groups contain chimeric haplotypes,
 11 single chromosomes that are themselves patchworks of *V. vinifera* and *V. labrusca* segments (Fig.
 12 2B). The presence of these chimeric structures provides the definitive genomic signature of a
 13 segmental allopolyploid, verifying that partially homologous chromosomes from these divergent
 14 species paired and exchanged genetic material (homoeologous exchange) during the breeding
 15 history. Together, these findings establish that Kyoho was shaped by both the differential
 16 segregation of whole chromosomes and extensive interspecific recombination.

17 To quantify the overall genomic contribution, we aggregated the ancestry assignments across the
 18 genome. The Kyoho genome is composed of approximately 71.2% *V. vinifera* and 28.8% *V.*
 19 *labrusca* ancestry, biasing towards the *V. vinifera* parent (Supplementary Fig. S5A and Table S8).
 20 This ancestry partitioning independently corroborated by SNPs/Indels/SVs analysis. We found that
 21 chromosomal regions assigned to a *V. labrusca* origin harbored a significantly higher density of
 22 private SNPs/Indels/SVs compared to *V. vinifera* regions. Given that the variants were called
 23 against PNT2T, this density disparity perfectly aligns with the greater evolutionary divergence of
 24 the *V. labrusca* segments, validating our ancestry model (Supplementary Fig. S5B).

25 We further dissected the sex-determining region (SDR) on chromosome 2 to specifically
 26 characterize the ancestral origin of this functional locus (Zhou *et al.*, 2017). The genetic
 27 architecture within this ~150 kb region across the different haplotypes revealed a clear divergence
 28 into two distinct groups, corresponding to the male/hermaphroditic (H) and female (F) forms (Fig.
 29 2C). By leveraging structural topology, we traced the specific parental lineages of these haplotypes.
 30 Notably, the SDR on Kyoho.hap1 was structurally homologous to the *V. labrusca* F haplotype (VL.

1 hap1), whereas Kyoho.hap4 corresponded to the *V. vinifera* F haplotype (PNT2T). Conversely,
2 both Kyoho.hap2 and Kyoho.hap3 aligned with the *V. labrusca* H haplotype (VL. hap2). These
3 results elucidate the hybrid composition of the SDR locus and confirm that Kyoho is a
4 hermaphrodite (HHFF) carrying both haplotype types.

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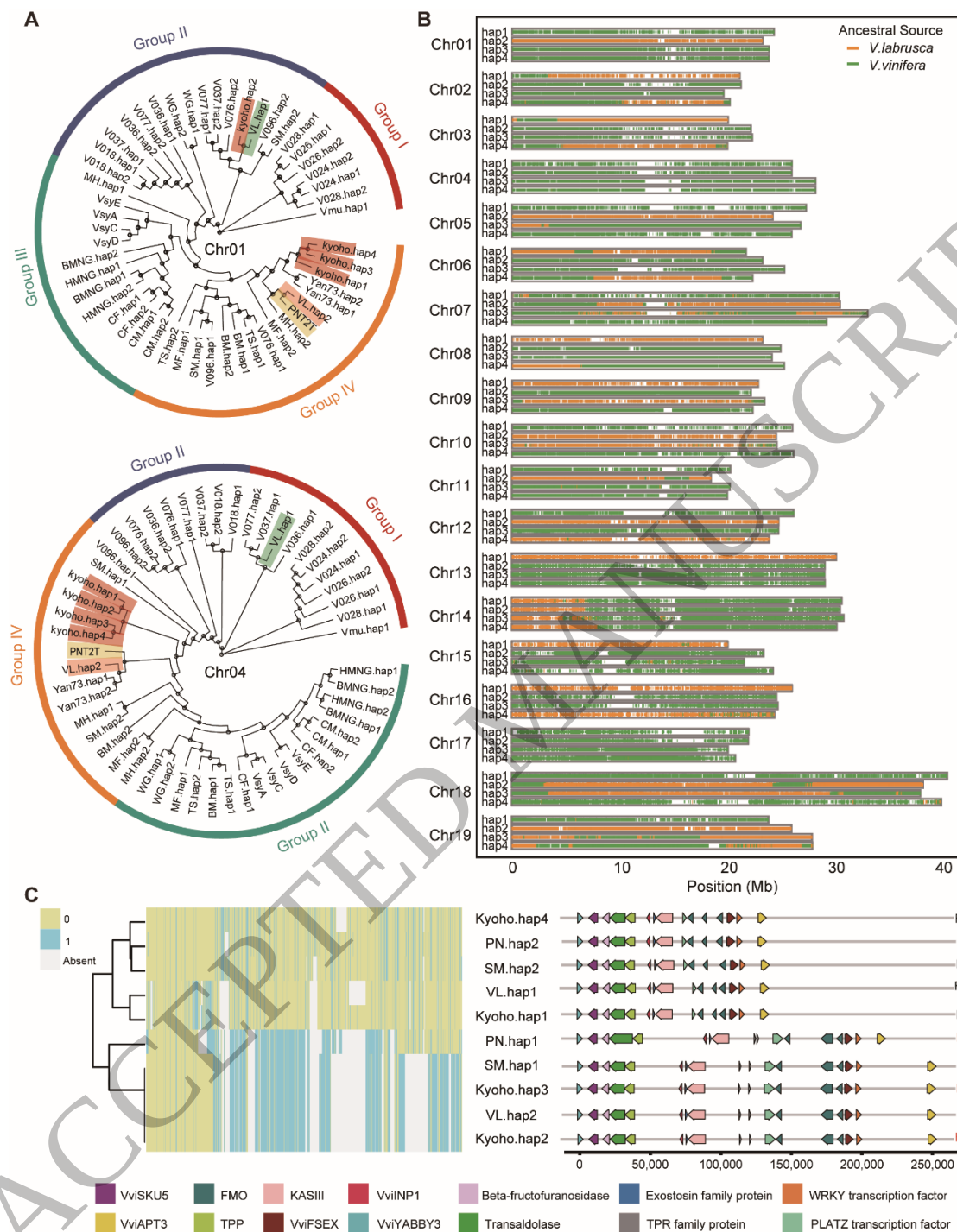


Figure 2. Evolutionary genomic origin and haplotypic composition of Kyoho. A. maximum likelihood phylogenetic tree using single-copy genes from chromosome Chr01 and Chr04 categorizes haplotypes into four groups (see other chromosomes in Figure S4). Group I: Approximate *V. rotundifolia*, Group II: Approximate *V. labrusca*, Group III: Approximate *V. vinifera*, and Group IV: the complete *V. vinifera* (the higher the proportion of single-copy genes

from *V. vinifera*, the closer it is to *V. vinifera*). B. Genomic landscape of ancestral segments across the 19 chromosomes of the four Kyoho haplotypes. Fragments inferred to originate from *V. labrusca* are colored orange, and those from *V. vinifera* are colored green, illustrating the mosaic composition of the genome. C. The sex-determination region in Kyoho haplotypes. The heatmap and dendrogram (left) compare this region across different Kyoho and reference haplotypes. The schematic diagram (right) illustrates the organization of candidate functional genes within this locus, linking specific haplotypes to either hermaphrodite (H) or female (F) phenotypes.

Transcriptomic landscape of the Kyoho mosaic genome

Having established the complex mosaic architecture of the Kyoho genome, we next sought to determine how this complex ancestry translates into functional gene expression. We generated comprehensive transcriptome profiles from leaf and three fruit developmental stages, identifying 20,480 homologous quadruplets (Figs. 3A, 3B; Supplementary Table S9). Focusing on 17,186 robustly expressed quadruplets, we categorized their regulatory patterns into 15 distinct motifs (Supplementary Fig. S6). The regulatory landscape was defined by striking local asymmetry; balanced expression was rare (3.0%), whereas asymmetric modes were predominated. These asymmetric patterns were primarily classified into three categories: single-haplotype dominance (42.7%), two-haplotype dominance (43.6%), and single-haplotype suppression (10.7%). Specifically, single-haplotype dominance was relatively evenly distributed among the four haplotypes (ranging from 9.7% to 11.3%), while two-haplotype dominance showed significant variation, with the hap1/hap2 combination being the most prevalent (13.9%). Collectively, these findings reveal a strong expression bias in Kyoho alleles, whereas haplotype-dominant or two-haplotype dominant patterns overwhelmingly surpass balanced expression. Functional enrichment revealed that specific dominant haploids drive distinct biological modules; for instance, hap1*Vitis*000012_hap2*Vitis*000206(*MADS-box*); hap1*Vitis*0076_hap3*Vitis*007617(*PSY*); hap2*Vitis*000205_hap4*Vitis*000209(*PL-ATZ*) preferentially regulate floral/embryonic development and light response pathways, respectively (Supplementary Table S10). Further analysis of the four homologous gene copies in the SDR region revealed no antagonistic effects between the genes from *V. labrusca* and *V. vinifera*; instead, both sets of genes were found to contribute positively to the observed traits (Supplementary Table S11).

To further evaluate whether there is any systematic deviation from a random, symmetrical relationship between *V. labrusca* and *V. vinifera* genomes in the Kyoho genome, we evaluated the competing influences of tissue identity versus genomic origin. We systematically compared 68,164 homologous gene pairs (representing pairwise alignments between *V. labrusca*-derived and *V. vinifera*-derived alleles) by integrating differentially expressed genes (DEGs) (Supplementary Fig. S7 and Table S12). To comprehensively investigate the regulatory divergence between *V. labrusca*- and *V. vinifera*-derived alleles, we classified the differentially expressed homologous pairs into eight distinct patterns based on their expression trajectories (D-D, U-U, D-N, N-D, U-N, N-U, D-U, U-D) (Supplementary Fig. S8). Strikingly, a substantial proportion, ranging from 48.7% to 75.8% across different comparisons, exhibited directionally inconsistent expression patterns (e.g., D-N, N-D, U-N, N-U, D-U, and U-D). This widespread prevalence of inconsistent regulatory responses indicates that allelic pairs do not strictly adhere to a synchronized expression program. Instead, they display significant plasticity, suggesting that the *V. labrusca* and *V. vinifera* alleles independently tune their contributions to meet specific developmental demands (Supplementary Table S13).

Finally, we quantified the stability of these regulatory patterns to understand their evolutionary implications. Among 17,750 homologous gene pairs exhibiting directional bias, only 4,086 (23.4%) showed constant bias (maintaining a fixed dominance direction across all tissues). The vast majority (13,227 pairs, 74.5%) displayed semi-constant bias (biased in some tissues but balanced or silent in others). This distribution confirms that allelic dominance in Kyoho is largely conditional and plastic rather than genetically fixed (Fig. 3C). We further explored this plasticity by analyzing 18,162 tissue-specifically expressed genes (TPGs), which Uniform Manifold Approximation and Projection (UMAP) dimensionality reduction resolved into four distinct co-regulatory archetypes (Supplementary Fig. S9). These genes were categorized into single-origin specific ('Vv' or 'Vl') and shared ('both') patterns (Fig. 3D). As shown in the expression quantification, while a subset of TPGs retains co-specificity ('both'), a significant proportion undergoes subfunctionalization, where expression is exclusively driven by either the *V. vinifera* or *V. labrusca* allele depending on the tissue. The heatmap trajectory (Fig. 3E) vividly illustrates these regulatory archetypes, ranging from strict co-enrichment to mutually exclusive tissue specificity. Collectively, these patterns demonstrate that tissue developmental context, rather than rigid genomic bias, is the primary driver of allelic divergence. By flexibly recruiting distinct ancestral

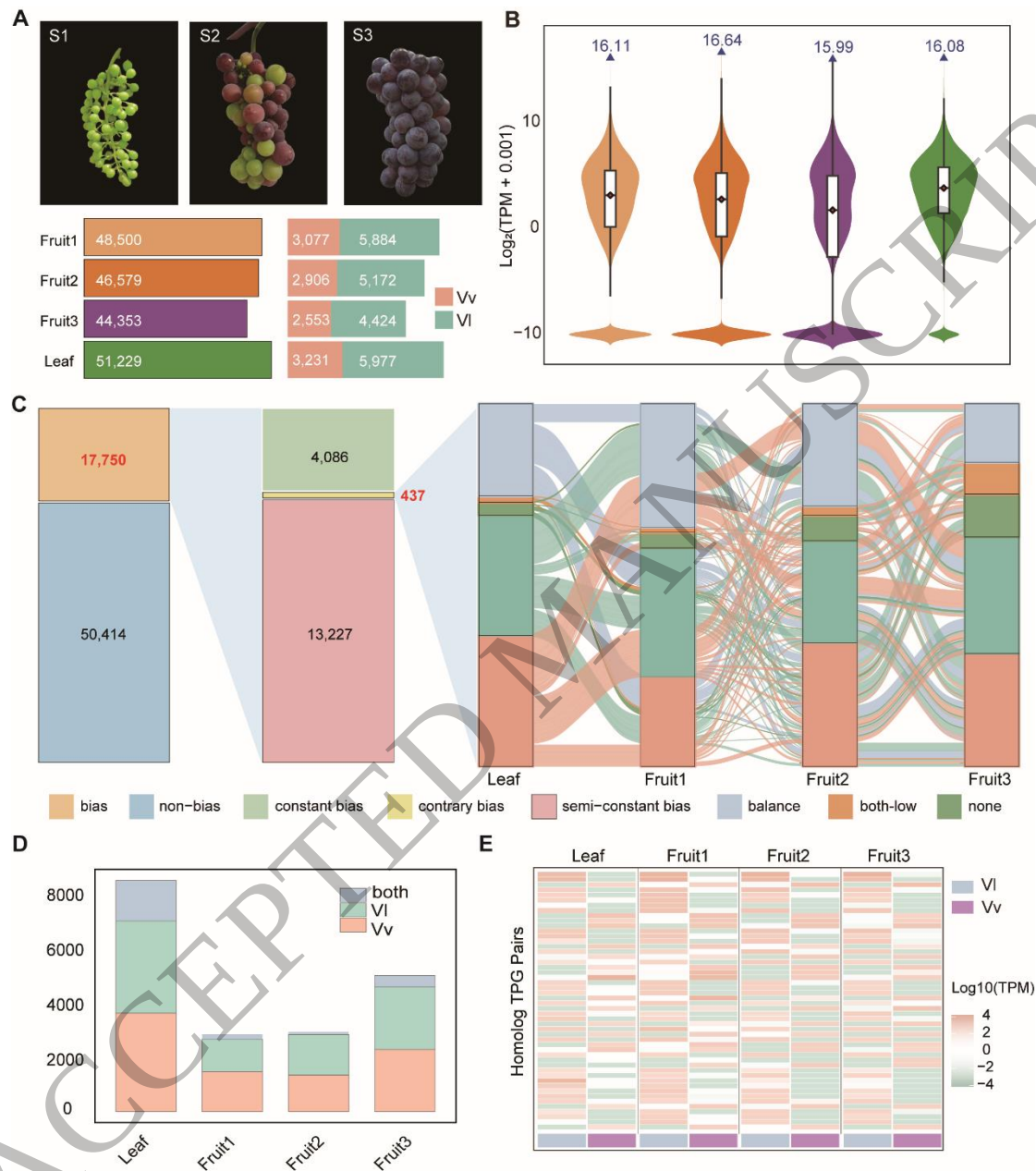


Figure 3. Parentalexpression bias in tetraploid Kyoho grape. A. Fruit at three developmental stages: S1 (young), S2 (veraison), and S3 (mature); The left panel shows the total number of expressed genes, and the right panel shows the number of genes with parental expression bias ('VI' and 'Vv' represent *V. labrusca*-bias and *V. vinifera*-bias, respectively). B. Violin plots of gene expression levels (log₂ TPM) and bar charts of expressed gene counts for each tissue and stage. C. Overall classification (left) and dynamic shifts (right) in parental expression bias across tissues,

visualized with an alluvial plot. The alluvial plot on the right systematically illustrates the dynamic transition trajectories of homologous gene pairs in terms of expression preference states during leaf and three fruit development stages, described by flow direction and bifurcation. Each flow line represents a gene pair and depicts its switching path between five possible expression states – no expression, both-low expression (both alleles), VL dominance, Vv dominance, or balanced. D. Tissue-specific homologous gene pair pattern statistics (‘VL’ = *V. labrusca* expression; ‘Vv’ = *V. vinifera* expression; ‘both’ = bilateral simultaneous expression). E. Heatmap of expression levels for the top 50 homologous gene pairs from the two ancestries across all samples.

DNA methylation landscape of the Kyoho mosaic genome

To elucidate the epigenetic landscape underpinning the Kyoho mosaic genome, we performed whole-genome bisulfite sequencing across leaf and three sequential fruit developmental stages (S1-S3). While CG and CHG methylation levels remained high and stable globally, the dynamic landscape was defined by a substantial increase in the CHH context during fruit maturation (Supplementary Fig. S10). The average CHH methylation levels rose more than five-fold, from 1.45% in the S1 stage to 7.65% in the S3 stage (Fig. 4B). Consistent with previous studies (Peng et al., 2025, Song et al., 2024b), TEs exhibited the highest average methylation (49.3% in CG context), significantly surpassing genic regions (29.97%) (Supplementary Fig. S10 and Table S14). Overall, this specific accumulation of CHH methylation signals a robust activation of the RNA-directed DNA methylation (RdDM) pathway, functioning as a global genomic stabilizer during the metabolically active ripening process (Supplementary Fig. S11).

We next examined whether the epigenetic landscape is symmetrical between the constituent ancestries (*V. vinifera* vs. *V. labrusca*). Weighted methylation analysis revealed a consistent quantitative bias: *V. vinifera*-derived origin displayed higher methylation levels than their *V. labrusca* counterparts across genome. This asymmetry was most pronounced in the CHH context within TEs. For instance, at the S3 stage, the mean CHH methylation level of *V. vinifera* TEs reached approximately 11.5%, distinctly surpassing the 9.8% observed in *V. labrusca* TEs (Supplementary Fig. S12 and Table S15). This divergence likely attributes to the differential TE loads accumulated during the distinct evolutionary histories of the two species. Crucially, however, this ancestral offset was subordinate to the overarching developmental program. As fruit development progressed, methylation levels in both ancestries increased synchronously, following

parallel upward trajectories V1 (0.009~0.676): Vv (0.010~0.670) (Supplementary Fig. S13). This finding indicates that the Kyoho genome employs a tightly coordinated regulatory system where developmental drive serves as the primary determinant, while genomic origin plays a subtle modulatory role.

To dissect the mechanism driving these patterns, we analyzed methylation distribution at high resolution. We found that methylation peaks, particularly in the CHH context, colocalized strongly with TE-dense regions regardless of their ancestral origin (Fig. 4A, Supplementary Fig. S14). At the level of individual homologous gene pairs, we uncovered a striking temporal dynamic: significant methylation divergence between *V. vinifera* and *V. labrusca* alleles was not constitutive but transient. Divergence peaked specifically during the fruit expansion stage (S2, $P < 0.05$) before re-converging (Fig. 4C). Comparative analysis across tissues and developmental stages classified gene pairs into two distinct categories: conserved pairs, which maintain consistent methylation states (such as BM-BM, IM-IM, and UM-UM), and remodeled pairs, which exhibit dynamic state transitions (including BM-IM, BM-UM, UM-BM, UM-IM, IM-BM, and IM-UM) (Fig. 4D, Supplementary Table S16). This stage-specific plasticity underscores a dynamic strategy of transient allelic divergence. By temporarily relaxing epigenetic synchrony to accommodate the rapid metabolic flux of expansion, the Kyoho genome effectively facilitates the targeted transcriptional subfunctionalization required for adaptive hybrid development.

Finally, to identify the functional targets of this epigenetic control, we examined genes located in hypermethylated islands. Enrichment analysis revealed a significant overrepresentation of gene families associated with genome defense (e.g., retrotransposon-derived proteins) and large, tandemly duplicated clusters such as receptor-like kinases (RLKs) and proteases (Supplementary Table S16). The hypermethylation of these loci is primarily driven by their proximity to or containment of TEs. This underscores a fundamental trade-off in the hybrid genome: DNA methylation acts as a genomic guardian, strictly silencing repetitive elements to maintain integrity, even if it means imposing a repressive epigenetic environment on adjacent multi-copy gene families.

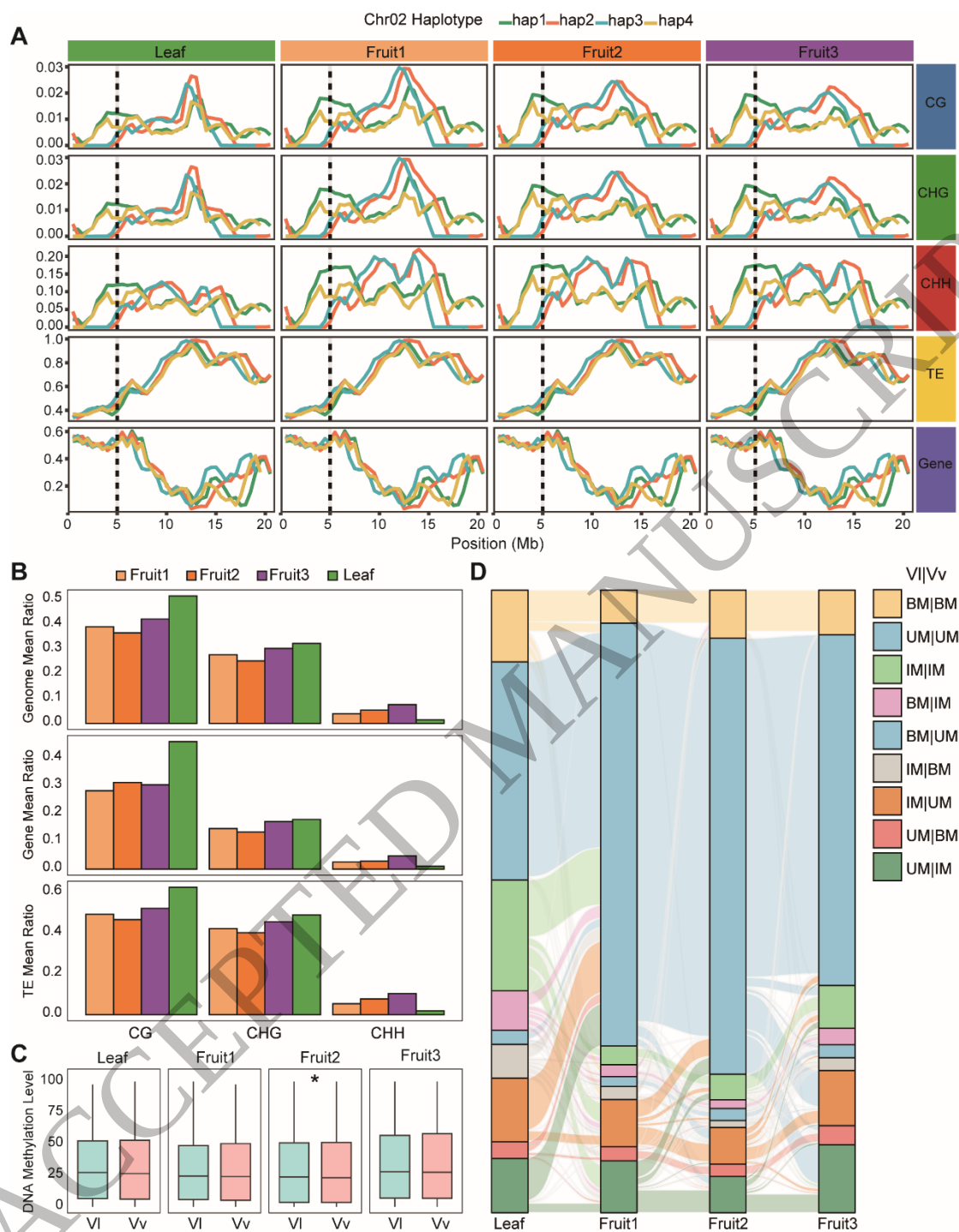


Figure 4. Allelic-specific DNA methylation dynamics and parental bias in Kyoho. A. Distribution of DNA methylation (in CG, CHG, and CHH contexts), transposable element (TE), and gene densities along chromosome 2 for the four phased copies across leaf and three fruit developmental stages. B. Average weighted methylation levels across the genome, gene bodies, and TEs. C. Comparison of gene body methylation between homologous alleles from the *V*.

labrusca (Vl) and *V. vinifera* (Vv). D. Alluvial plot showing the transitions of gene body methylation (gbM) states in homologous gene pairs across tissues and developmental stages. Gene body methylation was discretized into three states: Body Methylated (BM; highly methylated gene bodies), Intermediately Methylated (IM; moderate gbM levels), and Unmethylated (UM; low or negligible gbM). For each homologous pair, combined states include symmetric patterns, such as both methylated (BM|BM) and both unmethylated (UM|UM), as well as asymmetric configurations between the *V. labrusca* (Vl) and *V. vinifera* (Vv) ancestries (e.g., BM|UM, BM|IM, IM|UM).

Alternative splicing landscape of the Kyoho during fruit development

To capture the post-transcriptional complexity of the Kyoho genome, we integrated full-length Iso-Seq and short-read RNA-Seq data, establishing a comprehensive catalog of 82,750 alternative splicing (AS) events. Intron retention (RI) emerged as the dominant mode (67.1%), consistent with its prevalence in plants (Fig. 5A). The AS landscape exhibited a distinct temporal dynamic during fruit development. The transition from early fruit expansion (S1 to S2) involved moderate splicing modulation, including 2,510 differentially alternatively spliced (DAS) events. In contrast, the transition to ripening (S2 to S3) was characterized by a marked upregulation in splicing regulation, generating 3,820 DAS events (Fig. 5B). This intensification suggests that AS serves as a critical mechanism for proteomic diversification specifically during the complex metabolic shifts of ripening. Concurrently, isoform usage shifted toward exon inclusion during late development; notably, comparison of splicing inclusion difference (dPSI) revealed a significant increase in positive dPSI values for RI events in the S3 vs. S2 comparison (Fig 5C and 5E), indicating a stage-specific regulation of transcript processing.

We investigated the relationship between DAS events and differentially expressed genes (DEGs). The overlap between these two regulatory layers was limited to 11.9%, suggesting that AS operates largely as an independent regulatory mechanism rather than merely mirroring transcriptional changes. Correlation analysis between expression differences ($\log_2FC$) and inclusion level differences (dPSI) revealed no significant linear correlation ($r \approx 0$) across developmental stages (Fig. 5F). However, an inverse relationship was frequently observed where increased isoform inclusion coincided with gene downregulation (58.9%) (Fig. 5F), a pattern consistent with nonsense-mediated decay (NMD) feedback mechanisms. Furthermore, temporal analysis indicated

1 that splicing modulation often precedes steady-state changes in gene expression (Fig. 5G), positing
2 AS as a rapid response mechanism for fine-tuning the functional proteome prior to substantial
3 transcriptional reprogramming.

4 To determine how the mosaic ancestry influences post-transcriptional regulation, we analyzed
5 homologous alleles derived from *V. labrusca* and *V. vinifera* (Supplementary Fig. S15-17). The
6 distribution of AS event types was consistent across all four haplotypes, with RI remaining the
7 dominant category (Supplementary Fig. S15). Comparison of homologous gene pairs revealed a
8 dual regulatory strategy characterized by both evolutionary conservation and functional
9 divergence. On one hand, a high degree of conservation was evident in the splicing code; inclusion
10 levels (dPSI) between *V. vinifera* and *V. labrusca* homologs were strongly correlated (Pearson $r =$
11 0.766 , $P < 0.001$; Supplementary Fig. S16D). The majority of events showed identical splicing
12 modalities (e.g., RI in both ancestors), as indicated by the predominance of counts along the
13 diagonal in event-type combination matrices (Supplementary Fig. S16C and S16B).

14 On the other hand, substantial divergences were identified where homologs adopted distinct
15 splicing fates. Specific loci exhibited differential processing, such as RI in one allele pairing with
16 Exon Skipping (SE) or Alternative 3' Splice Site (A3) usage in its homolog (Supplementary Fig.
17 S17B). This divergence allows the hybrid genome to bypass diploid constraints, enabling one allele
18 to maintain the canonical isoform while the other explores novel splicing variants. Pathway
19 enrichment analysis suggested this divergent splicing is selectively recruited for key
20 developmental processes. For instance, allele-specific intron retention was observed in auxin
21 efflux carriers, and alternative 3' splice site usage was detected in GRF-interacting factors. These
22 findings indicate that differential processing of ancestral alleles contributes to the
23 subfunctionalization of genes regulating traits such as organ size and ripening timing
24 (Supplementary Table S17).

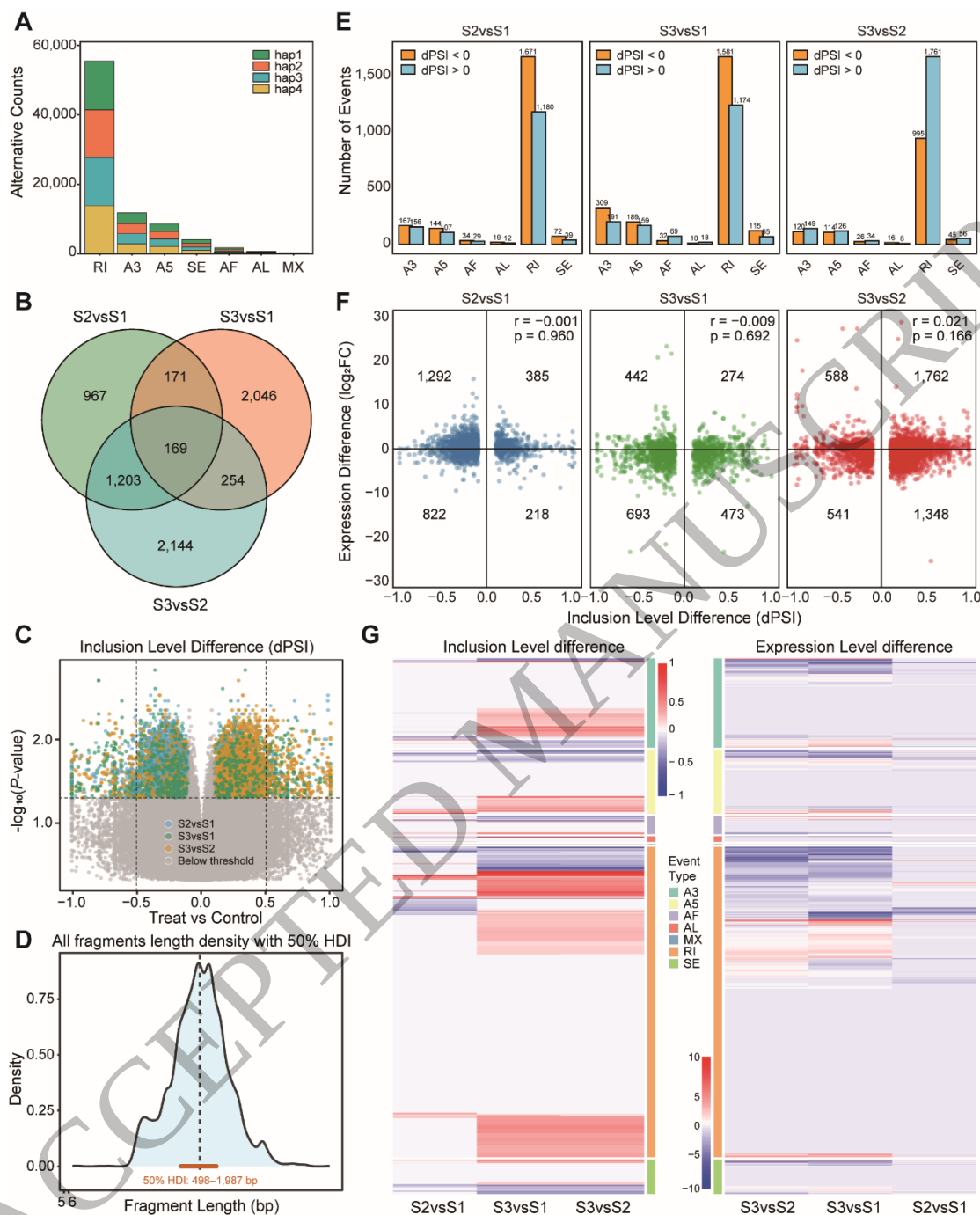


Figure 5. Iso-Seq and RNA-seq of alternative splicing dynamics during grapevine fruit development. A. Counts of alternative splicing (AS) event types identified across the four haplotypes. B. Venn diagram of differentially spliced (DAS) events overlapping between developmental stage comparisons. C. Volcano plot of differential AS events, showing the change in percent spliced-in (dPSI) versus statistical significance. D. Density plot of variable fragment lengths, indicating the 50% highest density interval (HDI). E. Number of significant DAS events

per type, showing increased ($dPSI > 0$) or decreased ($dPSI < 0$) inclusion for each stage comparison. F. Correlation between splicing inclusion difference ($dPSI$) and gene expression change ($\log_2FC$) for DAS events. G. Heatmaps of $dPSI$ values (left) and corresponding gene expression changes ($\log_2FC$, right) for significant DAS events, clustered by event type.

Integrative multi-omics reveal allele-specific gene regulation

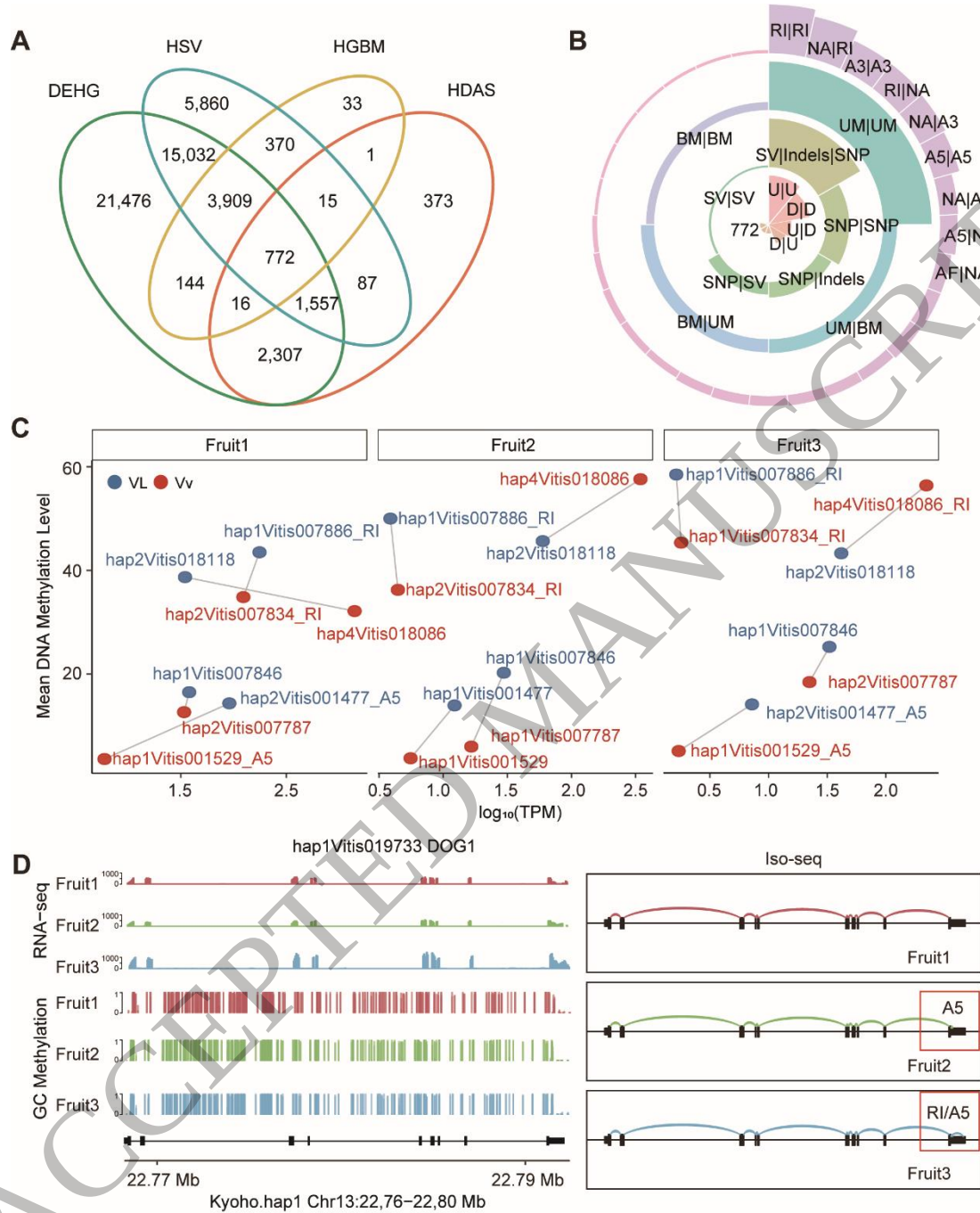
To dissect allele-specific gene regulation, we integrated multi-omics data covering genomic variants (SNPs/Indels/SVs), DNA methylation, transcriptomic changes, and alternative splicing (i.e., HSV, HGBM, DEHG, and HDAS) (Fig.6A). Structural variation emerged as a primary determinant of transcriptional abundance, evidenced by the substantial overlap of 15,032 genes exhibiting both HSV and DEHG. However, the regulatory landscape is not a simple binary output of genomic structure; rather, it exhibits distinct stratification. Transcriptional divergence appeared most plastic, with 21,476 genes showing unique expression changes (Unique DEHG) independent of large structural shifts, epigenetic modifications, or alternative splicing. Conversely, 5,860 genes harbored silent SVs (Unique HSV) that did not trigger downstream regulatory changes, highlighting the genome's capacity to buffer against structural perturbations.

At the intersection of these layers, we identified a core module of 772 genes that exhibited convergence across all four regulatory signals (HSV, DEHG, HGBM, and HDAS) (Fig. 6B and Fig. 6C). This multi-layered coordination suggests that for this specific gene set, structural differentiation anchors a synchronized regulatory recalibration to drive robust sub functionalization. Beyond this core set, we detected a significant group of 2,307 genes showing coordinated HDAS and DEHG changes in the absence of dynamic methylation (HGBM). Visual examination of specific candidates, including *DOG1* (*hap1Vitis019733_hap3Vitis019798*), a sugar transporter (*hap1Vitis023513_hap2Vitis023241*), Absciscic acid receptor (*hap4Vitis024149_hap2Vitis024109*), *ZEP* (*hap1Vitis020191_hap3Vitis020204*), *PIN* (*hap1Vitis011230_hap2Vitis011150*) and a *MYB* transcription factor (*hap1Vitis013173_hap2Vitis013123*), confirmed a strong mechanistic link between splicing events and homologous gene expression divergence (Fig. 6D; Supplementary Fig. 18 and Table S18). Despite this coordination, the splicing machinery retains the capacity for independent modulation. A distinct subset of genes (including 373 unique DAS events) exhibited differential splicing in the absence of other regulatory changes. This independent mode serves as a rapid response unit,

1 particularly during the transition to ripening, exemplified by isoform switching in *hap2Vitis018118*
2 (Fig. 6C), generating proteomic diversity without requiring the slower, energy-intensive overhaul
3 of the transcriptional baseline. In contrast, DNA methylation functioned primarily as a dependent
4 stabilizer; only 33 genes exhibited unique methylation differences, whereas 3,909 genes tracked
5 with structural and expression divergence.

6 Critically, while these loci lacked dynamic methylation changes, our analysis revealed that splicing
7 stability is intrinsically linked to the static methylation landscape. By analyzing data from different
8 fruit development stages, we focused on key regulatory mechanisms of fruit development, without
9 considering gene integration associated with ancestral structural variations. This structure is based
10 on the analysis of 6,415 differential splicing events, 50,537 DEGs, and 39,095 dynamic
11 methylation sites (Supplementary Fig. S19A and Table S19). We observed that UM were
12 significantly more prone to alternative splicing BM (Supplementary Fig. S19 B and Table S19).
13 The observed splicing events, such as RI and SE, preferentially mapped to hypomethylated regions
14 (Supplementary Fig. S19C). This correlation supports a model in which a stable low-methylation
15 environment ensures sufficient chromatin accessibility. Consequently, this allows the splicing
16 machinery to generate alternative isoforms, facilitating functional specialization in the absence of
17 dynamic epigenetic remodeling. In addition, functional enrichment, *HRGP*, *TBL*, *ZEP*, *PSY*,
18 *SPPase*, *GRAS*, and *Aux/IAA* are involved in a variety of key physiological processes, including
19 cell wall remodeling, pigment synthesis, energy metabolism, transcriptional regulation, and
20 hormone signal transduction, which together maintain the overall regulatory mechanism of fruit
21 development (Supplementary Fig. S20).

1 **Figure 6. Integrative multi-omics analysis of homologous gene pairs.** A. Venn diagram showing



2 the overlap of differential expressed homologous genes (DEHG), genes with differential
 3 alternative splicing (HDAS), homologous genes with differential gene-body methylation
 4 (HGBM), and homologous genes with structural variants (HSV). B. Sunburst plot illustrating
 5 multi-omic divergence between ancestral alleles, including alternative splicing, methylation
 6 (Display of homologous differential alternative splicing event types with counts greater than 5),
 7 SV, and expression. The concentric rings represent hierarchical expansion only, and categories in

different rings are not in a one-to-one correspondence. C. Gene expression, DAS and DNA methylation for homologous gene pairs from the q (blue, *V. labrusca*) and s (red, *V. vinifera*) subgenomes across fruit stages. D. Genome browser view of a candidate gene showing RNA-seq coverage, CG methylation, and Iso-Seq isoforms. An alternative 5' splice site (A5) or intron retention (RI) event is highlighted (red box).

Discussion

In this study, we generated a haplotype-resolved, near-complete genome for the tetraploid Kyoho grape, serving as a landmark resource to dissect regulatory logic of a complex segmental allopolyploid. Our integrative analysis highlights three key insights: (1) we quantitatively resolved its highly asymmetric, mosaic genomic architecture; (2) we overturned the assumption of global ancestral dominance—where alleles from one progenitor consistently out-express the other—revealing instead a flexible, tissue-adaptive regulatory model; (3) we established AS, rather DNA methylation, as the primary driver of functional divergence between ancestral alleles. Collectively, this work provides a new model for understanding how stability and plasticity are maintained in complex polyploid genomes under selective pressures, which will facilitate future polyploid breeding efforts.

Unlike classic allopolyploids with clearly demarcated subgenomes (Wang *et al.*, 2019, Zhang *et al.*, 2015, Hu *et al.*, 2019, Song *et al.*, 2024b, Zhou *et al.*, 2023)), Kyoho is a segmental allopolyploid shaped by extensive inter-ancestral recombination. Its homologous chromosomes are themselves chimeras of *V. vinifera* and *V. labrusca* ancestry. This structural mosaicism challenges the rigid subgenome-dominant models often applied in polyploid studies (OIV, 2017, Chalhoub *et al.*, 2014). Instead, the extensive recombination between ancestral genomes breaking linkage drag (Lynch & Force, 2000) and creating a heterogeneous substrate that necessitates the flexible, chimeric regulatory patterns we subsequently identified.

The functional output of this mosaic genome is governed by dynamic, developmental context-driven trade-offs rather than a static hierarchy. The vast majority of homologous gene pairs exhibiting overwhelming allele-specific expression bias (biased pairs) display variable dominance across tissues. This suggests that the observed expression patterns do not merely reflect an averaging of ancestral differences. Instead, artificial or natural selection has favored the specific ancestral alleles that are most advantageous for given physiological demands (Yoo *et al.*, 2013).

1 This independence is particularly critical during fruit maturation, where specific ancestral lineages
 2 confer distinct adaptive traits. By retaining conditional dominance, artificial and natural selection
 3 have shaped a regulatory landscape that provides both functional robustness and evolutionary
 4 plasticity through lineage-specific subfunctionalization (Lynch & Force, 2000).

5 Beyond transcriptional abundance, DNA methylation acts as a global synchronizer rather than a
 6 driver of ancestral divergence. Although *V. vinifera* regions exhibit higher baseline methylation
 7 due to differential TE loads (Matzke & Mosher, 2014), both ancestries undergo a synchronized
 8 surge in CHH methylation (RdDM) during ripening. This global reinforcement silences the chaotic
 9 TE landscape (Tang *et al.*, 2020, Huang *et al.*, 2019, Song *et al.*, 2024a), ensuring that the
 10 underlying structural asymmetries do not disrupt the developmental program. Crucially, a transient
 11 plasticity window at the S2 stage, where methylation divergence peaks before re-synchronizing,
 12 suggests that the epigenome serves as a dynamic filter. It maintains general genomic integrity while
 13 permitting brief regulatory relaxation to accommodate the rapid metabolic shifts of hybrid fruit
 14 development (Slotkin & Martienssen, 2007, Li *et al.*, 2023).

15 In contrast to the global synchrony of the methylome, AS is a rapid, orthogonal regulatory layer
 16 driving lineage-specific functional divergence (Reddy, 2007, Baralle & Giudice, 2017, Tognacca
 17 *et al.*, 2023, Reddy *et al.*, 2013, Alyahya & Taybi, 2024). Operating largely independently of
 18 transcriptional abundance, AS resolves a key regulatory puzzle: for thousands of loci lacking
 19 explanatory genomic variants or methylation differences, AS acts as the primary determinant for
 20 modulating allele-specific expression. Selection has favored this mechanism to fine-tune essential
 21 physiological pathways, evidenced by allele-specific splicing patterns in core ripening regulators,
 22 including *ZEP* (abscisic acid biosynthesis/ receptor), *Aux/IAA* (auxin signaling), *DOG1* (dormancy
 23 regulation), and *Myb* transcription factor (Guo *et al.*, 2021, Alyahya & Taybi, 2024, Hirschberg,
 24 2001). By leveraging these lineage-specific splicing signatures, the hybrid genome bypasses rigid
 25 DNA sequence constraints, enabling the precise recruitment of advantageous alleles to coordinate
 26 fruit maturation.

27 Importantly, the AS-driven regulatory strategy is not unique to grape. In other polyploids,
 28 subgenome- or allele-biased AS, alongside epigenetic remodeling, repeatedly emerges as a key
 29 driver of evolutionary divergence. For example, asymmetric AS profile shaped subgenome
 30 adaptation in hexaploid wheat and *Brassica napus* (Gao *et al.*, 2021, Yao *et al.*, 2020, Rousseau-

Gueutin *et al.*, 2020, de Jong & Adams, 2023, Wang & Wang, 2022). In parallel, polyploid tobacco and *Arabidopsis* illustrate extensive epigenomic and TE reconfiguration during adaptation (Mhiri *et al.*, 2019, Yuan *et al.*, 2020, Jiang *et al.*, 2021, Srikant *et al.*, 2024). Together, these studies indicate that allele-specific AS is a primary driver of functional divergence between ancestral alleles during development.

In conclusion, the haplotype-resolved, multi-omics Kyoho dataset provides a powerful toolkit to overcome long-standing challenges in polyploid crop improvement, facilitating the transition to haplotype-aware breeding. Our findings enable the development of markers tracking specific ancestral segments with desirable traits, such as the *V. labrusca*-derived disease resistance (Cadle-Davidson, 2008) or *V. vinifera*-derived flavor, across a complex recombination landscape. Moreover, the discovery that valuable alleles are mobilized via conditional plasticity implies that latent ancestral traits can be activated through targeted environmental or developmental cues. Ultimately, this work serves as a conceptual and practical blueprint for dissecting complex traits and accelerating genetic gain in segmental allopolyploid crops.

Materials and methods

Plant material and growth conditions

Kyoho grapevines were cultivated in the vineyards of the Agricultural Genomics Institute, Chinese Academy of Agricultural Sciences (Shenzhen, Guangzhou, China). For karyotyping, one-year-old dormant shoots were collected and propagated in 5x5 cm pots in a greenhouse at 26°C to stimulate the growth of active root meristems. For sequencing, samples of leaves, buds, roots, and fruits at three distinct developmental stages (S1: young green, S2: veraison/expansion, S3: ripening) were collected, flash-frozen in liquid nitrogen, and stored at -80°C until use.

Genomic DNA sequencing

High-molecular-weight (HMW) genomic DNA was isolated from a mixture of young leaves, buds, and roots using a modified CTAB/SDS phenol-chloroform extraction protocol combined with a Qiagen column-based purification step. The integrity and purity of the HMW DNA were confirmed before library construction. A PacBio HiFi library was constructed and sequenced on a PacBio Sequel II platform, generating 163.9 Gb of high-fidelity (HiFi) reads. An Oxford Nanopore Technology (ONT) library was prepared and sequenced on GridION X5 and PromethION

platforms, yielding 44.9 Gb of ultra-long reads. For chromatin conformation capture, fresh leaves were cross-linked with 2% formaldehyde, and a Hi-C library was prepared by digesting chromatin from isolated nuclei with the *DpnII* restriction endonuclease. The library was sequenced on an Illumina platform (PE150), producing approximately 50 Gb of paired-end reads.

Transcriptome and epigenome sequencing

Total RNA was extracted from leaves and the three fruit developmental stages (S1, S2, S3). For each sample, three biological replicates were utilized for short-read sequencing and two replicates for long-read sequencing. Two complementary transcriptome sequencing strategies were employed: (1) Illumina short-read RNA-seq libraries were prepared after rRNA depletion (Shanghai Personalbio, China) and sequenced on an Illumina platform, yielding ~12 Gb per library. (2) PacBio full-length cDNA (Iso-Seq) libraries were prepared (Hope Bio, China) and sequenced on a PacBio Sequel II platform, producing a total of ~400 Gb of subreads. For DNA methylation profiling, whole-genome bisulfite sequencing (WGBS) was performed on the same tissues and stages (two biological replicates each). Libraries were constructed and sequenced by Shanghai Personal Biotechnology Co., Ltd. (Shanghai, China) on an Illumina platform, yielding approximately 96 Gb of raw data.

Genome assembly and annotation

To assemble a haplotype-resolved, near-T2T reference genome, we employed a hierarchical strategy. First, the genome size and heterozygosity were estimated from HiFi reads by analyzing the 21-mer frequency distribution with Jellyfish (v2.3.0) (Marcais *et al.*, 2018) and GenomeScope (v2.0) (Ranallo-Benavidez *et al.*, 2020). Then, a haplotype-resolved assembly was generated using hifiasm (v0.19.7-r598) (Cheng *et al.*, 2021) with HiFi and ONT reads, using Hi-C data to phase the four haplotypes (--n-hap 4). The resulting contigs were anchored to 76 pseudo-chromosomes based on synteny to the PNT2T reference genome using RagTag (v2.1.0) (Alonge *et al.*, 2022). Chromosome-scale scaffolding was further refined using the Hi-C data with Juicer (v1.6) (Durand *et al.*, 2016) and 3D-DNA (v190716) (Dudchenko *et al.*, 2017), followed by manual curation of the Hi-C contact maps in Juicebox (v2.17.00) (Durand *et al.*, 2016). Remaining gaps were manually filled by aligning HiFi and ONT reads with Minimap2 (v2.24-r1122) (Li, 2018) and inspecting alignments in the Integrative Genomics Viewer (IGV) (v2.13.1) (Robinson *et al.*, 2023). The final assembly quality was assessed with Seqkit (v2.2.0) (Shen *et al.*, 2016), and gene

completeness was evaluated with BUSCO (v5.3.0) (Simao *et al.*, 2015) against the embryophyta_odb10 database. “hap 1–4” are arbitrary technical assemblies used for quality assessment (like BUSCO) and do not represent biologically linked whole-genome haplotypes.

TEs were annotated de novo using a combination of Repeat Modeler/RepeatMasker (Flynn *et al.*, 2020) and Extensive de-novo TE Annotator (EDTA) (v1.9.6) (Ou *et al.*, 2019), with LTR elements further refined using LTR_retriever (v2.9.0) (Ou & Jiang, 2018). Centromeric regions were identified by locating dense arrays of a 107 bp tandem repeat with TRF (v4.09) (Benson, 1999), and telomeric repeats ((TTAGGG)_n) were identified with the Telomere Identification Toolkit (tidk) (v0.2.0) (Brown *et al.*, 2025). Protein-coding genes were predicted using a combination of ab initio, homology-based, and transcriptome-assisted approaches.

Predicted protein-coding genes were functionally annotated using InterProScan (v5.48-83.0) (Blum *et al.*, 2025) to assign GO terms, InterPro domains, and Pfam families. Additionally, protein sequences were searched against the UniProt Swiss-Prot database, and functional descriptions were transferred from the best-scoring hits.

Evolutionary genomics and ancestry analysis

To assess genomic diversity and identify structural variants, we performed whole-genome alignments between the four Kyoho haplotypes and the PNT2T reference using Nucmer (v4.0.0) (Marçais *et al.*, 2018) (--maxmatch -c 100 -b 500 -l 50). SVs, including inversions, translocations, and duplications, were identified from the filtered alignments using SyRI (v1.6.4) (Goel *et al.*, 2019, Goel & Schneeberger, 2022) and visualized with plotsr (v1.0.1) (Goel & Schneeberger, 2022). SNPs, Indels, and large-effect SVs (>50 bp) were extracted from the SyRI output for each haplotype, and shared versus private variants were identified using bcftools.

To determine the ancestral origin of the Kyoho genome, a phylogenomic analysis was conducted. Single-copy orthologs were identified across 31 *Vitis* genomes (Supplementary Table S20) using OrthoFinder (v2.5.5) (Emms & Kelly, 2019). Conserved blocks from single-copy gene alignments were extracted with Gblocks (v0.91b) (Castresana, 2000), and the optimal amino acid substitution model (JTT+I+G+F) was determined using ProtTest3 (v3.4.2) (Darriba *et al.*, 2011). Phylogenetic trees were constructed with IQ-TREE (v8.2.12) to infer the ancestral contributions of *V. vinifera* and *V. labrusca* to the Kyoho genome. To further resolve genome-scale relationships, we

performed whole-genome alignments of each component pair in VL/PNT2 using nucmer and calculated genetic distances with distmat (<http://www.bioinformatics.nl/cgi-bin/emboss/distmat>). For the Kyohogenome, we then applied Jukes–Cantor correction to non-overlapping 50 kb windows.

Transcriptomic analyses

To quantify allelic-specific expression and analyze homologous gene expression patterns, raw Illumina RNA-seq reads were filtered for quality using fastp (v0.23.2) (Chen *et al.*, 2018). Clean reads were aligned to each of the four haplotype-resolved genome assemblies separately using HISAT2 (v2.2.1) (Kim *et al.*, 2015), and gene-level read counts were quantified with featureCounts (v2.0.8) (Liao *et al.*, 2014).

We defined homologous relationships at two levels: four-way tetrads and pairwise ancestral homologs. First, to define homologous tetrads (1:1:1:1), an all-vs-all BLASTP search (E -value $< 1e^{-5}$) was performed across the four phased copies. Sets of four genes exhibiting reciprocal best-hit relationships across all pairwise comparisons were retained. For expression motif analysis within tetrads, TPM values were normalized to the total expression of the tetrad following established methods (Ramirez-Gonzalez *et al.*, 2018), and each tetrad was assigned to one of 15 ideal expression categories based on minimal Euclidean distance.

Second, to evaluate divergence between ancestral lineages, we identified homologous gene pairs between *V. labrusca* and *V. vinifera*. A BLASTP search was performed using the *V. labrusca*-derived proteins as the query and *V. vinifera*-derived proteins as the subject. High-confidence pairs were filtered using stringent criteria: E -value $< 1 \times 10^{-10}$, identity $> 95\%$, and alignment coverage $\approx 100\%$.

To identify differentially and tissue-specifically expressed genes, gene counts were analyzed with DESeq2 (v1.38.3) (Love *et al.*, 2014). Genes were classified as upregulated (U) or downregulated (D) if they met the significance thresholds (adjusted $P < 0.05$ and $|\log_2 \text{FoldChange}| > 1$; otherwise, they were classified as non-significant (N).

We systematically compared regulatory modes between ancestral homologs by combining their individual differential states across pairwise tissue comparisons. This resulted in eight regulatory modes: coordinated (U-U, D-D), divergent (U-D, D-U), and unilateral (U-N, N-U, D-N, N-D).

Pairs classifying as stable in both (N-N) were excluded from the divergence analysis. Tissue-specificity was further assessed using the TAU algorithm, with genes exhibiting a tissue-specificity index (TSI) ≥ 0.85 defined as tissue-specific.

To assess allele-specific expression bias between *V. labrusca* (C_{Lab}) and *V. vinifera* (C_{Vin}) homologs within each tissue, we applied a strict joint-threshold strategy. Homologous pairs with low abundance (average TPM ≤ 0.5 in both lineages) were labeled "BothLow" and excluded. For expressed pairs, bias was assessed using a binomial test under the null hypothesis $P = 0.5$ (where $n = C_{Lab} + C_{Vin}$), with P-values corrected using the Benjamini-Hochberg (FDR) procedure.

Significant expression bias was declared only if a pair met two criteria: (1) a False Discovery Rate (FDR) ≤ 0.01 ; and (2) an absolute fold-change magnitude $|\log_2 C_{Lab} / C_{Vin}| > 1$ for both raw counts and TPM values (calculated with a pseudocount to smooth variance). The direction of bias was determined by the sign of the log fold-change. Pairs not meeting these criteria were classified as "Balanced." Finally, the stability of this bias was categorized across the four tissues into four classes: Constant bias (consistent direction across all significant contexts), Semi-constant bias (consistent direction in some contexts, mixing with Balanced/BothLow), contrary bias (direction shifts between contexts), and No bias.

Epigenomic analyses

To analyze DNA methylation dynamics, raw WGBS reads were quality-trimmed using Trimmomatic (v0.39) (Bolger *et al.*, 2014). Clean reads were aligned to the four haplotype genomes using Bismark (v0.23.1) (Krueger & Andrews, 2011) with the Bowtie2 (v2.1.0) (Langmead & Salzberg, 2012) aligner. PCR duplicates were removed with deduplicate_bismark, and methylation calls were extracted using bismark_methylation_extractor. Only cytosine sites with a minimum coverage of 2x were retained for analysis. Methylation levels were calculated for CG, CHG, and CHH contexts. The distribution of methylation across genomic features (genes, TEs, promoters) and 500 kb chromosomal bins was analyzed, and metaplots were generated using deepTools (v3.5.6) (Ramirez *et al.*, 2014).

To systematically compare gene body methylation (gbM) levels between the *V. labrusca* and *V. vinifera* ancestral lineages across tissues, we harmonized the methylation records for the four haplotypes (hap1-hap4) and mapped them to the identified homologous gene pairs of *V. labrusca*

and *V. vinifera* ancestries. The analysis encompassed four conditions (leaf and three fruit developmental stages), each with two biological replicates. We retained homologous pairs that possessed valid methylation data for both lineages in at least one replicate, calculating the mean methylation level across replicates to obtain paired within-tissue observations. Differences in gbM between *V. labrusca* and *V. vinifera* homologs were assessed using a two-sided paired t-test, with significance stratified at $P < 0.05$. Finally, to characterize the spatiotemporal dynamics and stability of the epigenetic landscape, we categorized genes into a three-state model: Body Methylated (BM), Intermediately Methylated (IM), and Unmethylated (UM). This analysis highlighted the regulatory divergence between the *V. labrusca* and *V. vinifera* genomic components.

Alternative splicing analyses

To create a comprehensive reference transcriptome for splicing analysis, raw PacBio Iso-Seq subreads were processed into circular consensus sequences (CCSs) using isoseq3 (v4.0.0) (<https://github.com/PacificBiosciences/IsoSeq/blob/master/iseq-clustering.md>). Full-length non-chimeric (FLNC) reads were identified, clustered, and polished to generate high-quality isoforms (accuracy $\geq 99\%$). These isoforms were further error-corrected using Illumina short-read data with LoRDEC (v0.9) (Salmela & Rivals, 2014). Corrected, non-redundant transcripts were mapped to the Kyoho genome using pbmm2 (v1.17.0) (<https://github.com/PacificBiosciences/pbmm2>) and collapsed using isoseq collapse to remove 5' end degradation artifacts. A comprehensive transcriptome annotation was generated by merging the Iso-Seq transcripts with the short-read based annotations using gffcompare (v0.11.2) (Pertea & Pertea, 2020).

To identify and quantify DAS, the SUPPA2 pipeline was used. This final annotation served as the reference for AS analysis with SUPPA2 (v2.3) (Alamancos *et al.*, 2015). Transcript abundances were quantified using Salmon (v1.10.3) (Patro *et al.*, 2017), and the percent spliced-in (PSI) value for each AS event was calculated. Differential alternative splicing (DAS) events were identified between conditions using a threshold of $|dPSI| > 0.1$ and an FDR < 0.05 . To evaluate the relationship between splicing and transcription, DAS results were integrated with differential gene expression data using normalized gene identifiers. For intersecting genes, global associations were estimated using Pearson and Spearman correlations. Directional consistency was assessed by categorizing

genes into quadrants based on the signs of dPSI and log2FoldChange (Q1/Q3: consistent; Q2/Q4: inconsistent).

For orthologous analysis, we adopted a one-to-one *V. vinifera* and *V. labrusca* homology framework. Standardized DAS events were compared across three intra-tissue and three inter-tissue conditions. Pairwise event type combinations flanking homologous gene pairs were enumerated to characterize the total load and compositional spectrum. Splicing conservation was assessed by correlating dPSI values and quantifying effect size differences $dPSI = dPSI_{vin} - dPSI_{lab}$. Using a threshold of $|dPSI| \geq 0.1$, pairwise regulatory patterns were defined as identical, opposite, or unilateral to examine regulatory symmetry.

Finally, structure characterization was analyzed for RI, A5, A3, AF, AL, and SE event. Event lengths were defined as the distance between endpoints (for RI, SE) or alternative splice sites (for A3, A5, AF, AL). Length distributions were modeled using kernel density estimation. We reported median, and interquartile range (IQR), and 50% Highest Density Intervals (HDIs) on both linear and log1p scales to accommodate heavy-tailed distributions.

Statistics and reproducibility

Statistical comparisons of continuous variables between groups were performed using the two-sided Wilcoxon signed-rank test, and differences in proportions were assessed using the two-sided Pearson chi-square test. All statistical analyses were conducted in R (v4.5).

Accession numbers

Iso-Seq, BS-seq and RNA-seq data have been deposited in the NCBI Sequence Read Archive under project number PRJNA1372951 and the National Genomics Data Center (NGDC) Genome Sequence Archive (GSA) (<https://ngdc.cnbc.ac.cn/gsa/>), with BioProject number PRJCA052795. The haplotype-resolved genome assemblies of Kyoho have been deposited in the National Center for Biotechnology Information (NCBI) under the BioProject accession PRJNA1372984.

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Author contributions

Y.Z., W.L. and C.L. conceived and designed the study. L.S. collected the plant materials. Yingchun Z. and X.W. assembled and annotated the haplotype-resolved genomes. Yingchun Z., Y.X. and C.C. performed grape photography. Yingchun Z., Z.J., X.F. and Q.F. analyzed the evolutionary genomic origins. Yingchun Z., J.Y., Q.L. and T.Z. conducted transcriptome analysis. Yingchun Z., Y.S., S.Y., Z.A. and R.Y. performed DNA methylation analysis. Y.Z., Sheng Y., Y.D., L.Z., and L.T. conducted Iso-seq transcriptome analysis. Yingchun Z. and Y.P. wrote the manuscript. All authors provided critical feedback and approved the final manuscript.

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Data availability

The haplotype-resolved genome assemblies of Kyoho have been deposited in Zendo (<https://doi.org/10.5281/zenodo.18047369>). The complete genome assembly and gap-filling workflow, along with sequencing data, are available on our lab's GitHub repository at GitHub@zhouyflab.

Competing interests

The authors declare no competing interests.

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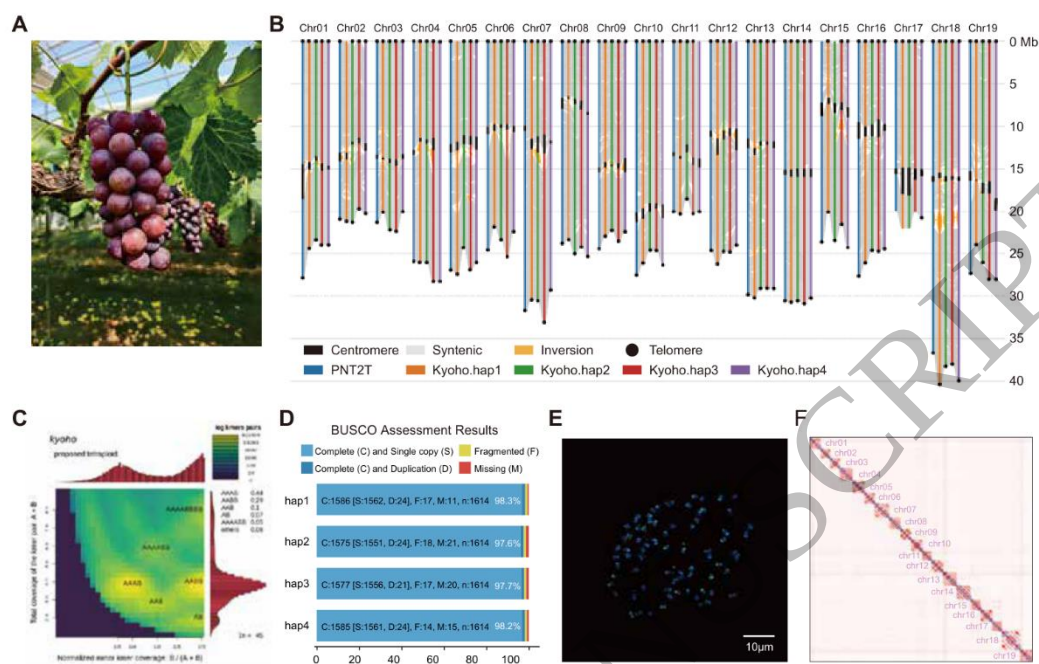


Figure 1
210x297 mm (x DPI)

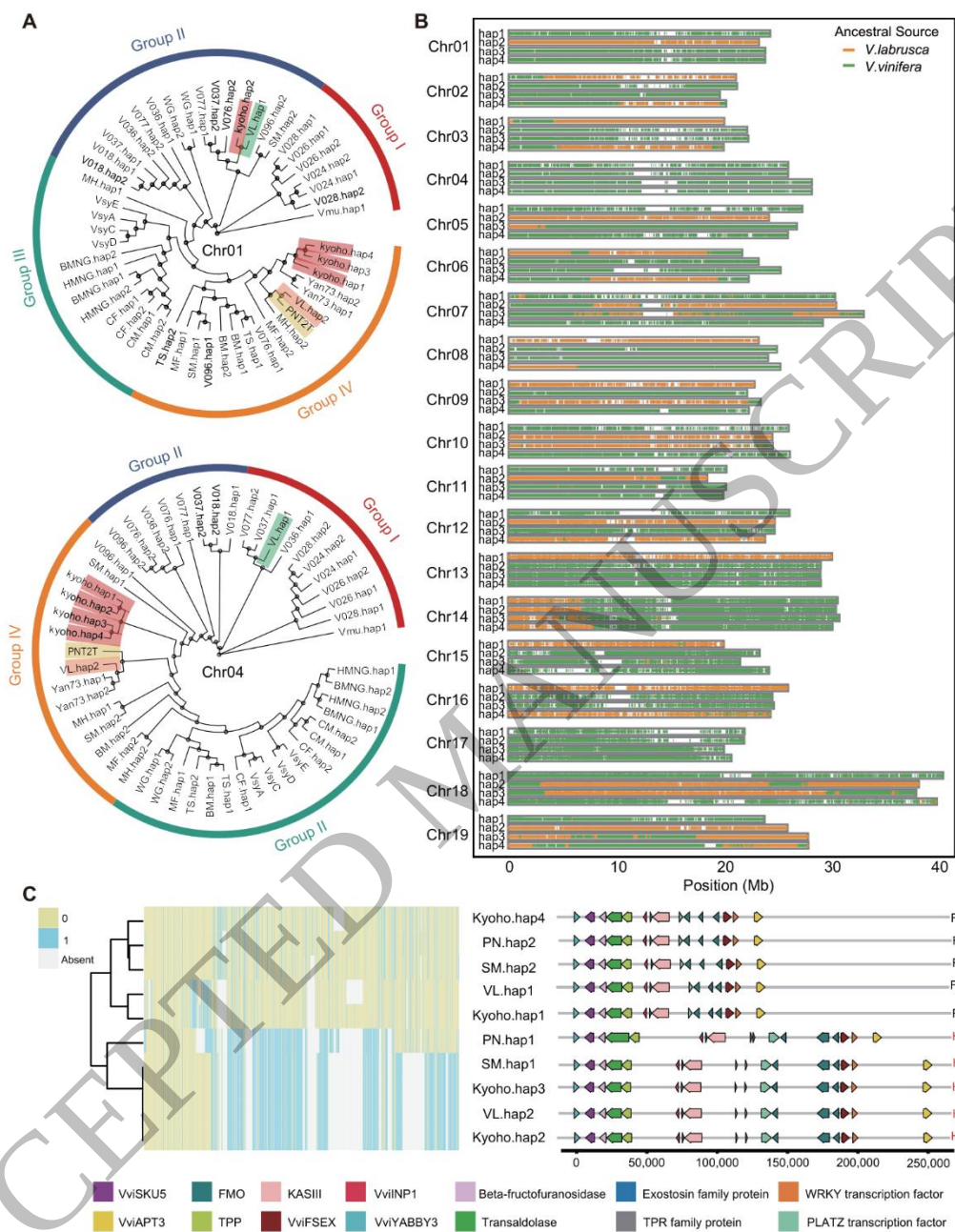


Figure 2
210x297 mm (x DPI)

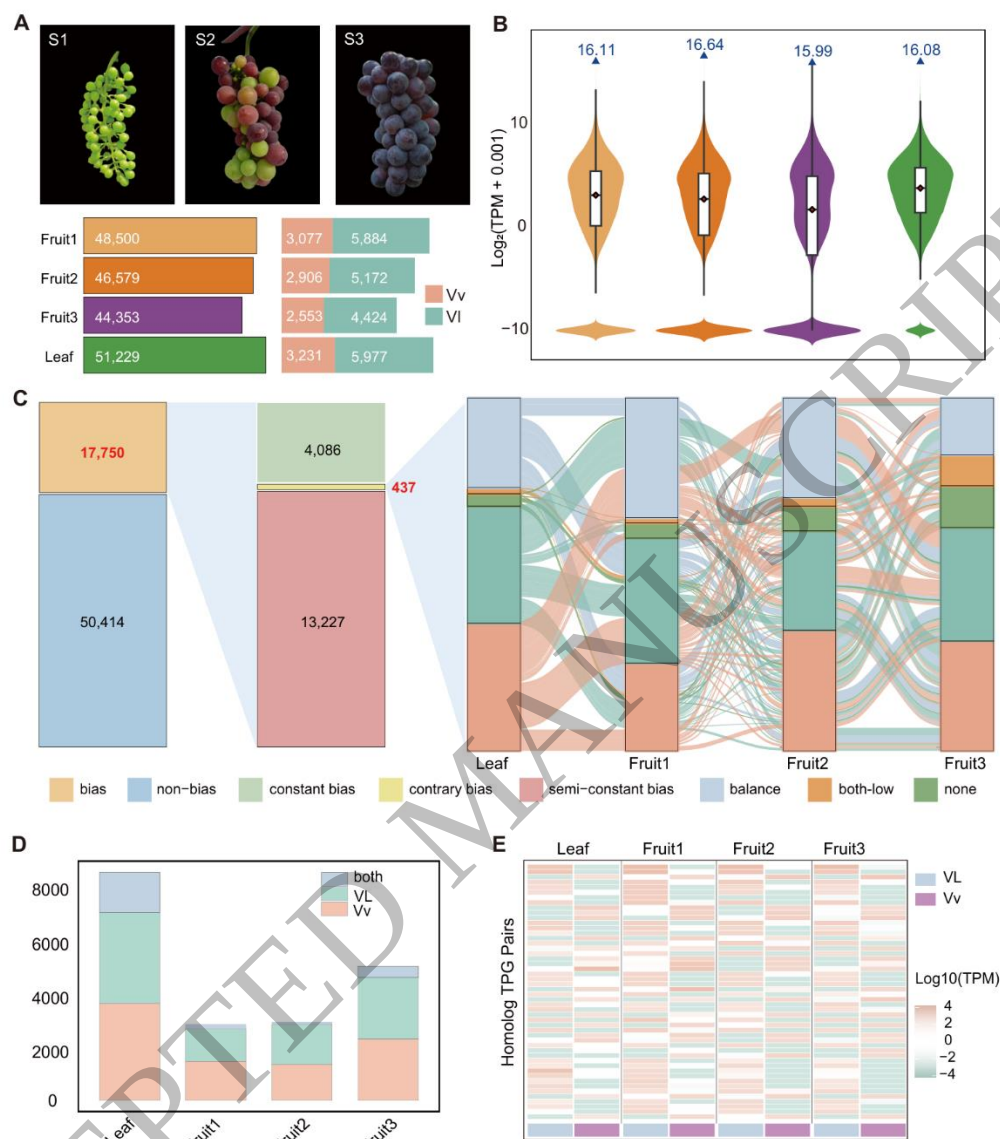


Figure 3
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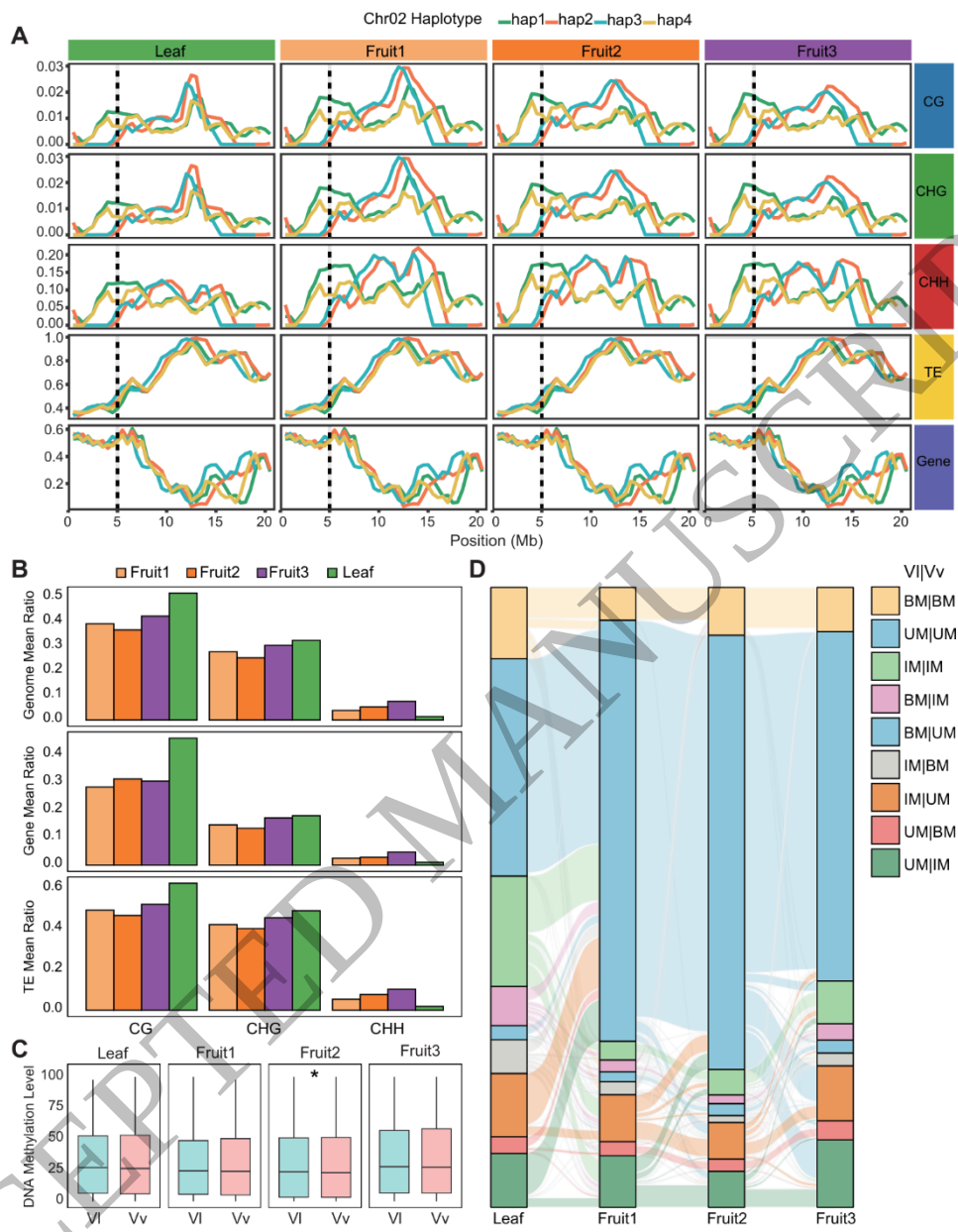


Figure 4
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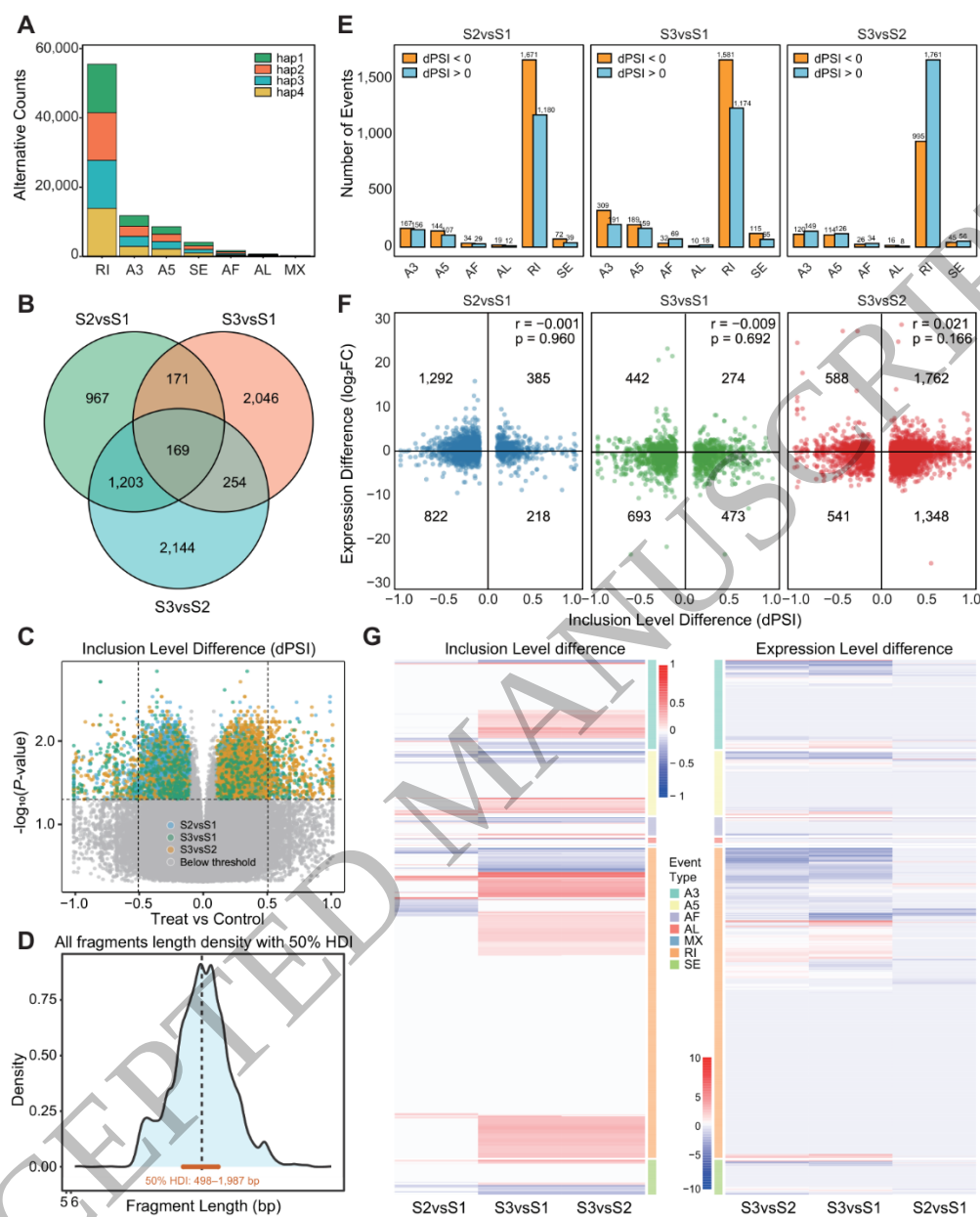


Figure 5
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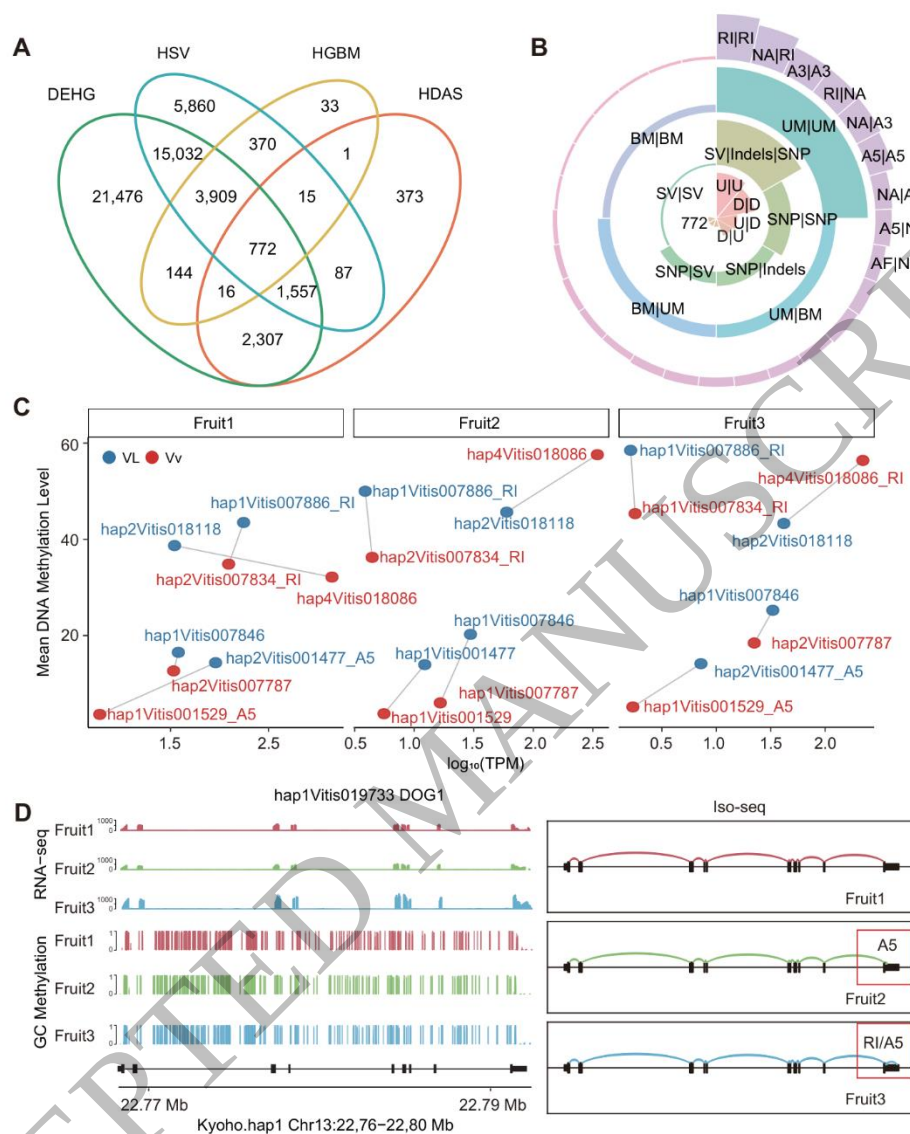


Figure 6
210x297 mm (x DPI)