

# Integrative genomics reveals the polygenic basis of seedlessness in grapevine

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## 27 Abstract

28 Seedlessness is a crucial quality trait in table grape (*Vitis vinifera* L.) breeding.  
 29 However, the development of seeds involved intricate regulations, while the  
 30 polygenic basis of seed abortion remains unclear. Here, we combine comparative  
 31 genomics, population genetics, quantitative genetics, and integrative genomics to  
 32 unravel the evolution and polygenic basis of seedlessness in grapes. We generated  
 33 four haplotype-resolved telomere-to-telomere (T2T) genomes for two seedless grape  
 34 cultivars, ‘Thompson Seedless’ (TS, syn. ‘Sultania’) and ‘Black Monukka’ (BM).  
 35 Comparative genomics identified a ~4.25 Mb hemizygous inversion on Chr10  
 36 specific in seedless cultivars, with seedless-associated genes *VvTT16* and *VvSUS2*  
 37 located at breakpoints. Population genomic analyses of 548 grapevine accessions  
 38 revealed two distinct clusters of seedless cultivars, tracing the origin of the  
 39 seedlessness trait back to ‘Sultania’. Introgression, rather than convergent selection,  
 40 shaped the evolutionary history of seedlessness in grape improvement. Genome-wide  
 41 association study (GWAS) analysis identified 110 quantitative trait loci (QTLs)  
 42 associated with 634 candidate genes, including novel candidate genes, such as three  
 43 *IIS GLOBULIN SEED STORAGE PROTEIN* and two *CYTOCHROME P450* genes,  
 44 and well-known genes like *VviAGL11*. Integrative genomic analyses resulted in 339  
 45 core candidate genes categorized into 13 groups related to seed development.  
 46 Machine learning based genomic selection achieved a remarkable 99% precision in  
 47 predicting grapevine seedlessness. Our findings highlight the polygenic nature of  
 48 seedless and provide novel candidate genes for molecular genetics and an effective  
 49 prediction for seedlessness in grape genomic breeding.

50

51 **Keywords:** Viticulture, Grape breeding, Seed abortion, Structural variations,  
 52 Genomic breeding, Genomic selection



## 54 Introduction

55 The production of seedless fruits leads to tremendous success in the global fruit  
 56 market<sup>1</sup>, such as bananas<sup>2,3</sup>, citrus<sup>4,5</sup>, watermelons<sup>6,7</sup>, and table grapes<sup>8</sup>. Seed abortion  
 57 in table grape has been a major focus of breeding efforts for decades, as seedless  
 58 grapes are highly preferred by consumers owing to improved tastes and convenience.  
 59 There are two primary methods employed to obtain seedless grapes. One involving  
 60 the application of phytohormones, applying gibberellin acid (GA) and cytokinin  
 61 analogs before the full bloom stage can effectively induce seed abortion in seeded  
 62 grapes<sup>9-11</sup>. Although this process has already become a common practice for  
 63 producing seedless grapes, it raises concerns about food safety and labor costs<sup>12</sup>.  
 64 Another one is based on genetic breeding of seedless grape cultivars. Breeders have  
 65 explored diploid and triploid breeding approaches and the embryo rescue strategy to  
 66 obtain new seedless varieties in recent decades<sup>13-16</sup>.

67 The development of various seed tissues in grapes involves intricate genomic  
 68 regulations. Previous studies have identified specific genes associated with different  
 69 tissues of seed development<sup>17</sup>. For instance, the formation of the seed coat  
 70 (integument) has been affected by genes like *VviAGL1*<sup>18-21</sup>, *VvMADS28*<sup>22</sup>,  
 71 *VviINO*<sup>23,24</sup>, and *VvHB63*<sup>25-27</sup>. Nutrient storage in the endosperm is controlled by  
 72 genes such as 7S and 11S globulin-like seed storage proteins<sup>28-30</sup>, while normal  
 73 embryo growth relies on several gibberellin (GA) genes<sup>31,32</sup>. Moreover, the growth of  
 74 ovules (young seeds) is influenced by genes such as *VvMADS39*<sup>33</sup>, *VvMADS45*<sup>34</sup>,  
 75 *VviABCG20*<sup>35-37</sup>, *VvFUS3*<sup>38</sup>, *VvNAC26*<sup>39</sup>, *VvβVPE*<sup>40,41</sup>, *VviASNI*<sup>42</sup>. In general,  
 76 multiple genes involved in tissue development of grapevine seeds. Seed abortion in  
 77 grapes can occur when any of the seed tissues fail to develop properly. However, the  
 78 polygenic basis of seedlessness in grapes remains unclear.

79 Previous studies have mapped multiple QTLs in different progenies in grapevine. In  
 80 the linkage map of 'Dominga' and 'Autumn Seedless', three QTLs for seed number

(SN) and six QTLs for seed fresh weight (SFW) were detected<sup>43</sup>. Similarly, in the ‘Muscat of Alexandria’ and ‘Crimson Seedless’ progeny, six QTLs for SN and ten QTLs for SFW were detected<sup>44</sup>. Recent comparative analyses, encompassing 28 different grape varieties (13 seeded and 15 seedless), have detected 34 candidate genes associated with the divergence between seeded and seedless lineages<sup>45</sup>. However, the restricted genetic background in progenies and limited population samples hinders the investigation of the polygenic basis of seedlessness in grapes.

In this study, we first generated haplotype-resolved T2T genomes for two seedless grapes, ‘Thompson Seedless’ (TS) and ‘Black Monukka’ (BM), and we compared these haplotype genomes with 11 other grape genomes to detect structural variations (SVs) between seeded and seedless genomes. Population genetic analysis was conducted on whole-genome sequencing (WGS) of 548 accessions to investigate the evolutionary history of seedlessness, while quantitative genetic analysis involved 444 accessions to map QTLs and key genes associated with seedlessness. Integrative genomic analysis incorporated three transcriptomes with 14 development stages, homologous genes related to 34 Gene Ontology (GO) terms, and 451 family genes and seven molecular markers previously reported with significant effects on seed development processes. Finally, genomic selection, based on polygenic model and machine learning algorithms, were applied in predicting the seedlessness trait in grapes. Collectively, we aimed to address five sets of questions. First, at genome level, how do the seedless cultivars compare with cultivars with seeds? Can we detect big SVs related to seedless/seeded cultivars? Second, what evolutionary factors has driven the origin of seedlessness during grape improvement? i.e., introgression or convergent selection? Third, based on large natural populations, can we map genetic loci and candidate genes involved in seed abortion in table grapes? Fourth, can we integrate genomic analyses to identify core candidate genes underlying seed abortion? Finally, can we employ machine learning based genome selection to enhance prediction precision in grape breeding? Overall, our work contributes to improving

109 the understanding of the polygenic basis of seedlessness and facilitate genomic  
110 breeding of grapes.

# 111 Results

## 112 Comparative genomics between seeded and seedless cultivars

113 To study the genetic basis of seedlessness, we generated haplotype-resolved T2T  
 114 assemblies for two seedless cultivars: TS and BM (**Fig. 1b, c**), utilizing high-depth  
 115 PacBio HiFi sequencing (~120× coverage) and Hi-C sequencing (~116× coverage;  
 116 **Extended Data Fig. 2c**). The evaluation of K-mer heterozygosity in the TS and BM  
 117 genomes, based on HiFi data, measured 1.51% and 1.41%, respectively. The quality  
 118 of these genomes meets the assessment standards of the T2T level<sup>46</sup>, with all  
 119 centromeres regions and mostly telomeres regions marked (**Fig. 1a and**  
 120 **Supplementary Table 1, 3**). Statistical analysis of variants revealed that TS and BM,  
 121 between their two haplotype genomes, harbor 5.35 Mb and 5.04 Mb of SNPs, 5.01  
 122 Mb and 4.54 Mb of insertions and deletions (InDels, < 50 bp), and 33.42 Mb and  
 123 31.87 Mb of SVs (≥ 50 bp), respectively (**Supplementary Table 4**). A unique  
 124 heterozygous inversion region (PN\_T2T, Chr15: 10.7-12.0 Mb) specific to the TS  
 125 hap2 genome was detected when aligning the four haplotypic genomes to PN\_T2T,  
 126 and several genes were found near the inversion breakpoints, such as *AGAMOUS*  
 127 (*VvAG2*), *AGAMOUS-LIKE 62* (*VvAGL62*), *OIL BODY-ASSOCIATED PROTEIN 2B*  
 128 (*VvOBAP2B*), and *GDSL esterase/lipase At1g29670*, which are involved in stamen  
 129 and carpel determining, early endosperm development, and oil body synthesis  
 130 (**Extended Data Fig. 3b and Supplementary Table 5**). The differential chromosomes  
 131 between the two haplotype genomes explains the polymorphism of alleles and the  
 132 variation in the number of genes (**Supplementary Table 1**).

133 To further investigate the SVs associated with seedless and seeded grapes, we aligned  
 134 a total of 15 genomes, including the five seedless genomes and ten seeded genomes,  
 135 to the PN\_T2T (**Extended Data Fig. 3a**). We detected a heterozygous inversion  
 136 (PN\_T2T, Chr10: 23.8-25.4 Mb) in seedless grape varieties (**Fig. 1d**). The  
 137 authenticity of these inversions was confirmed by Hi-C heatmaps and IGV<sup>47</sup> (**Fig. 1e**

138 **and Extended Data Fig. 4).** A total of 210 genes (Chr10: 21.75-26.00 Mb) and 237  
139 genes (Chr10:23.00-27.50 Mb) were identified in the inversion regions of TS hap1  
140 and BM hap1, respectively (**Fig. 1f and Supplementary Table 7-8**). Three seed  
141 development-related genes, *TRANSPARENT TESTA 16/ ARABIDOPSIS BSISTER*  
142 (*TT16/ABS*) and two *SUCROSE SYNTHASE 2 (SUS2)* genes, were discovered near  
143 the breakpoints of inversion region of the haplotype genomes. *TT16/ABS* controls the  
144 formation of the maternal-derived endothelial cells by interacting with *AGL11/*  
145 *SEEDSTICK (STK)* in the previous studies<sup>48-51</sup>. *VvTT16* was found to be present in  
146 both TS and BM haplotype genomes, while the two *VvSUS2* tandem duplication  
147 genes were hemizygous, present only in the hap1 genome of TS and BM  
148 (**Supplementary Table 6-7**). These findings suggest that the power of comparative  
149 genomics of haplotype-resolved T2T genomes in uncovering overlooked new  
150 candidate genes underlying crucial agronomic traits.

## 151 **Introgression rather than convergent evolution underlying the** 152 **evolvment of seedlessness in grapevine**

153 To explore the evolutionary history of seedlessness in grape improvement, we used  
154 WGS data from 548 grapevine accessions, including 46 seedless grapes, for  
155 population genetic analysis (**Supplementary Table 8**). A total of 4,462,797 SNPs,  
156 443,812 InDels, 487,204 SVs were identified by aligning WGS data to ‘Cabernet  
157 Sauvignon’ (CS) genome<sup>52</sup>. The phylogenetic tree split into six primary population  
158 branches: European wild grapes (*V. vinifera* ssp. *sylvestris* EU population, EU, n = 69),  
159 Middle East and Caucasus region wild grapes (*V. vinifera* ssp. *sylvestris* ME  
160 population, ME, n = 23), domesticated grapes (*V. vinifera* ssp. *vinifera*, VV, n = 352),  
161 American fox grapes (*V. labrusca*, VL, n = 5), hybrid of VV and VL grapes (VV×VL,  
162 *V. vinifera* × *Vitis labrusca*, n = 92), and outgroup grapes (OG, n = 7; **Fig. 2c and**  
163 **Extended Data Fig. 5**). The results showed two independent lineages of seedless  
164 grapes nested in the VV×VL and VV branches, respectively, which is also supported



165 by PCA (**Fig. 2b, c**). The seedless grapes nested in two branches, which could be  
166 driven by convergent artificial selection or introgression during grape improvement.

167 To distinguish convergent selection and introgression in generating seedless traits in  
168 different grapevine lineages, we performed the introgression analyses, throughout the  
169 whole genome using  $f_d$  statistics<sup>53</sup>, following previous studies<sup>54,55</sup>. Interestingly, we  
170 detected significant genomic signals of introgression at seedless associated locus (see  
171 the GWAS section) between VV and VV×VL seedless grapes (the upper 5th  
172 percentile,  $f_d = 0.266$ ,  $P = 1.28e-39$ ), including the redefined *SEED DEVELOPMENT*  
173 *INHIBITOR* (SDInew, 30.36-31.86 Mb) locus<sup>20</sup> and the newly detected QTL on Chr07  
174 (SDI2, 8.85-8.86 Mb; **Fig. 3c, d**). These results suggested that introgression rather  
175 than convergent artificial selection has driven the evolutionary history of seedlessness  
176 in grapes.

177 To validate the genetic relationship among seedless varieties, a network was  
178 constructed comprising 46 seedless grapes (35 VV and 11 VV×VL) based on the  
179 results of Identity-by-Descent (IBD) analysis (**Supplementary Table 8**). The result  
180 revealed that TS group and BM group serve as bridges for gene flow between VV and  
181 VV×VL clusters (**Fig. 2d, e**). In fact, BM grape traces its ancestry back to ‘Sultania’  
182 (TS) and ‘Ichkimar’<sup>56</sup>, with an IBD score of 0.50 supporting this observation  
183 (**Supplementary Table 9**). Additionally, we identified three grapes belonging to the  
184 ‘Sultania’ somatic variants or synonym group (‘Jingfeng seedless’, TS1, and TS2,  
185  $IBD > 0.95$ ), seven grapes classified as parent-offspring relationship ( $IBD > 0.50$ ),  
186 and 32 grapes ( $0.50 > IBD > 0.09$ ) that were closely connected to ‘Sultania’ variety  
187 (**Supplementary Table 9**). These findings provide evidence that ‘Sultania’ had been  
188 extensively employed in crossbreeding with local grape varieties to enhance quality  
189 and develop new seedless grape cultivars<sup>8,57</sup>. Furthermore, the seed abortion could  
190 also be caused by cytoplasmic male sterility (CMS)<sup>58</sup>. We analyzed the chloroplast  
191 and mitochondrial genomic variation and found nuclear inheritance, rather than CMS  
192 inheritance, played a crucial role in controlling seed abortion (**Extended Data Fig. 6**).

193 These findings corroborated IBD results that frequent introgression has facilitated the  
194 formation of seedlessness the VV and VV×VL populations (**Fig. 3e**). As a result, the  
195 origin of the seedlessness trait could be traced back to ‘Sultania’, and continuous  
196 introgression, rather than convergent evolution, led to seed abortion in seedless grape  
197 varieties.

## 198 **Genome-wide Association Study for Seed Abortion Trait**

199 To detect the QTLs and genes associated with seed abortion, we used three population  
200 for GWAS analysis, considering the effects of population structure in GWAS analyses:  
201 VV (35 seedless and 317 seeded grapes), VV×VL (11 seedless and 81 seeded grapes),  
202 and an admixed population (46 seedless and 398 seeded grapes; **Supplementary**  
203 **Table 8**). We identified a total of 110 QTLs (634 genes), including 20 QTLs (126  
204 genes) specific to the VV population, 18 QTLs (106 genes) specific to the VV×VL  
205 population, and 72 consensus QTLs (402 genes) in admixed population, respectively  
206 (**Extended Data Fig. 7 and Supplementary Table 10, 11**). GO analysis revealed that  
207 genes specific to VV×VL were enriched in defense response and lignin catabolic  
208 process ( $P < 0.05$ ), while genes specific to VV population were enriched in embryo  
209 development ending in seed dormancy and xylan metabolic process ( $P < 0.05$ ;  
210 **Extended Data Fig. 8 and Supplementary Table 12**).

211 Remarkably, two consensus regions exhibited high consistence in three populations:  
212 Chr07: 8.85-8.86 Mb and Chr18: 29.40-35.54 Mb (**Fig. 3a, b**). In Chr07 locus (SDI2),  
213 two genes, *REVERSE TRANSCRIPTASE ZINC-BINDING DOMAIN-CONTAINING*  
214 *PROTEIN (LOC104880636, Vitvi011893)* and *STRUCTURAL MAINTENANCE OF*  
215 *CHROMOSOMES PROTEIN (SMC1, Vitvi011891)*, were positioned within a tightly  
216 linked region with high linkage disequilibrium (LD) values (**Fig. 3c**). The  
217 *LOC104880636* gene is annotated as the regulation of seed growth on UniProt, and  
218 the *smc1* mutants produced arrested early embryo development and blocked  
219 cellularization of the endosperm<sup>59</sup> (**Supplementary Table 11**). Additionally, within

the 50 kb upstream region of SDI2, we identified a closely linked cluster of three tandem-duplicated genes, *IIS GLOBULIN SEED STORAGE PROTEIN*<sup>28-30,60</sup> (**Fig. 3c**), and designated them as *IIS globulin G1*, *G2*, and *G3* based on their genomic positions. Notably, two nonsynonymous mutations and one deletion related to seedlessness were detected across 14 grape genomes (**Fig. 4a and Extended Data Fig. 9**). Among them, both the heterozygous Asp-to-Val and Leu-to-Val mutations were specific in seedless grapes, except for the somatic mutations of ‘Black Corinth’ (BC) seeded grapes. However, the heterozygous deletion of the 18 amino acids in *IIS globulin G3* was only detected in seedless grapes (**Fig. 4a**). The relative expression values of these genes showed a significant correlation with seed phenotypes, especially from 40-50 days after flowering (DAF; **Fig. 4b**).

Another consensus region is located on the Chr18: 29.40-35.54 Mb, encompassing 17 QTLs and the reported SDI locus (Chr18: 29.83-31.34 Mb; **Fig. 3d**). Due to the narrow genetic background of the samples used in the previous study<sup>20</sup>, we redefined the SDI locus (SDInew, Chr18: 30.36-31.86 Mb) through GWAS analysis, revealing a total of eight QTLs (**Fig. 3d and Extended Data Fig. 7**). In this region, the population differentiation was lower than the genomic background between VV ( $n = 35$ ) and VV×VL ( $n = 11$ ) seedless grapes, as showed by the fixation indices ( $F_{ST}$ ) results, suggesting a relatively close genetic distance between the two populations (SDInew  $F_{ST} = 0.073$  vs. genome-wide  $F_{ST} = 0.126$ ; **Supplementary Table 13**). This finding was also supported by genetic diversity ( $\pi$ ) statistics. The two seedless populations showed similar genetic diversity on the SDInew locus (**Fig. 3d**). The  $f_d$  statistics<sup>53</sup> revealed numerous sites showing evidence of introgressions ( $f_d = 0.266$ ,  $P = 1.28e-39$ ), including numerous genes related to seedlessness, such as *CYP716A94*, *CYP716A17*, *VviAGL11*, etc. (**Fig. 3d**). Notably, several SNPs and InDels were highly associated with these candidate genes, especially in the promoter and coding sequence region (**Fig. 4c, d and Extended Data Fig. 10**), while several published molecular markers for seedlessness prediction, including e7\_VviAGL11<sup>44</sup>, 5U\_VviAGL11<sup>44</sup>,

P3\_VviAGL11<sup>18</sup>, and VMC7f2<sup>61</sup>, showed low predictive accuracy due to the absence of significant genotyping quality (**Fig. 4d and Supplementary Table 14**). We selected the top 10% of associated sites based on  $-\log_{10}[P]$  values within the SDInew locus (**Fig. 4e**), and the most promising site for seedlessness prediction is Chr18\_30874059, exhibiting a predicted accuracy of 97.8% for seedless grapes and 94.22% for seeded grapes in natural population.

Interestingly, we also detected two genomic regions for specific to each population (**Extended Data Fig. 7**). In the VV population, a specific region Chr01: 17.85-20.42 Mb harbored 13 QTLs and 67 genes, including seven primary genes involved in seed development, such as *NON-SPECIFIC LIPID TRANSFER PROTEIN GPI-ANCHORED 1 (LTGP1)*, *ARABIDOPSIS HISTIDINE KINASE 3 (AHK3)*, *B3 DOMAIN-CONTAINING TRANSCRIPTION FACTOR FUS3 (FUS3)*, *XYLOGLUCAN ENDOTRANSGLUCOSYLASE PROTEIN (XTH)*, *11-BETA-HYDROXYSTEROID DEHYDROGENASE B (SOP3)*, as well as previously reported VvMADS4<sup>62</sup> and VvARF2-1<sup>63</sup> (**Fig. 3a**). In the VV×VL population, a specific region Chr18: 15.14-20.57 Mb harbored eight QTLs and 33 associated genes. Among these genes, *SERINE DECARBOXYLASE 1 (SDC1)*, *ABC TRANSPORTER G FAMILY MEMBER 22 ABCG22* or VvPNWBC22.2(TANG, 2018 #20), *SUS2*, *LACCASE-14 (LAC14)* and *TT10/LAC15* were highly associated with seed development (**Fig. 3a**). Our results suggest that seed abortion can be regulated by the collaborative effects of multiple genes, and the mapped candidate genes and variable sites hold valuable potential applications in seedless grapes breeding.

## Integrative Genomic Analysis Identified 339 Seedless Candidate Genes

To elucidate the polygenic basis of seed abortion, we further utilized an integrative genomic analysis using transcriptomic analyses, seed development associated GO term genes, previously reported family genes and molecular markers, and GWAS

275 candidate genes, to identify the core candidate genes associated with seed abortion  
 276 (**Supplementary Table 17**). Among these, three transcriptomic groups, including 76  
 277 samples and 14 time points, were employed to detect differentially expressed genes  
 278 (DEGs) between seeded and seedless grapes at each development stage  
 279 (**Supplementary Table 15**). For the ‘Italia’ and ‘Hongju Seedless’ (HS) groups, we  
 280 identified a total of 2,680 significantly upregulated genes and 1,835 significantly  
 281 downregulated genes from the six time points (**Extended Data Fig. 13c**). Similarly, in  
 282 the ‘Pinot Noir’ (PN) and TS groups, we identified a total of 3,969 significantly  
 283 upregulated genes and 2,695 significantly downregulated genes (**Extended Data Fig.**  
 284 **13c**). ‘Himrod Seedless’ (Himrod) and ‘Jinzao Wuhe’ (Jinzao) were used serve as a  
 285 control for cross-validation during four fruit development stage. Interestingly, we  
 286 found *VviAGLII* was exclusively in the downregulated DEGs in the PN and TS  
 287 comparison, but not in the ‘Italia’ and HS comparison (**Extended Data Fig. 13a, b**).  
 288 In addition, 1,301 core upregulated genes and 616 core downregulated genes were  
 289 only identified in transcriptomic analyses using integrative genomic analysis, such as  
 290 *DORMANCY-ASSOCIATED PROTEIN HOMOLOG 4* (*DRM1 homolog 4*),  
 291 *NON-SPECIFIC LIPID-TRANSFER PROTEIN 2* (*LTP2*), *VICILIN-LIKE SEED*  
 292 *STORAGE PROTEIN*, *7S SEED STORAGE PROTEIN* (*7S GLOBULIN*), *TT10*,  
 293 *LAC17*, etc. (**Fig. 5a and Supplementary Table 11**).

294 To include more important genes involved in seed development, we performed a  
 295 protein sequence similarity alignment for all genes associated with 34 GO terms  
 296 related to seed development against the PN\_T2T reference genome (**Extended Data**  
 297 **Fig. 12a, b**). This yielded 6,529 homologous genes (see Methods), with 5,061 genes  
 298 only present in the GO pathway, including the *TT16/FBP24* gene identified in the  
 299 comparative genomics (**Fig. 5a**). Notably, the intersection between the RNA-seq  
 300 related genes and the GO homologous genes revealed 163 downregulated genes and  
 301 294 upregulated genes (**Extended Data Fig. 13b**), such as *LTP*, *11S GLOBULIN*,  
 302 *OLEOSIN*, *CELLULOSE SYNTHASE A CATALYTIC SUBUNIT* (*CESA4*, *CESA7* and

303 *CESA8*), etc. (**Supplementary Table 11**). The consensus GWAS genes and GO  
 304 homologous genes exhibited nine candidate genes, such as previously mentioned  
 305 *LOC104880636* (**Fig. 3c, 5a**). Furthermore, we integrated a total of 451 family genes  
 306 and seven molecular markers related to seed abortion from previous studies  
 307 (**Supplementary Table 16**). All these elements were aligned against the PN\_T2T  
 308 reference genome (e-value < 0.1) and exhibited overlap with candidate genes in other  
 309 analyses, including previously reported genes such as *VvβVPE*<sup>40</sup>, *HD-ZIP PROTEINS*  
 310 *ATHB-1/HAT5*, *ATHB-12/VvHB56* and *ATHB40/VvHB18*<sup>25</sup>, *VviASN1*<sup>42</sup>, *VvMJE1*<sup>64</sup>,  
 311 *VvLECI*<sup>65</sup>, *VvMADS2/VvSEPI*<sup>66</sup>, etc. (**Extended Data Fig. 13d and Supplementary**  
 312 **Table 11**).

313 Through integrative genomic analysis, we screened 339 core candidate genes,  
 314 categorized into 13 groups, by condition-based filtering that exhibited significant  
 315 differential expression between seedless and seeded grape cultivars (see Methods,  
 316 **Supplementary Table 11**). Among them, 77 genes were directly associated with seed  
 317 development-related GO homologous genes, and three groups deserve our attention:  
 318 Firstly, the differential expression of candidate genes in the endosperm development  
 319 impacts nutrient storage in seedless grapes (**Fig. 5b**). Nutrient deficiency could be  
 320 primary factor leading to seed abortion in later embryo development; Secondly, genes  
 321 involved in the regulation of lignin and cellulose synthesis/degradation in seed coat  
 322 exhibit higher activity levels in seeded grapes (**Fig. 5c**); Thirdly, candidate genes  
 323 manage the synthesis and transport of oil bodies ensuring efficient lipid accumulation  
 324 and utilization during seed development (**Fig. 5d**). Overall, our findings indicate that  
 325 multiple genes have an accumulative effect in the process of seed abortion, resulting  
 326 varying degrees of seedlessness<sup>67</sup>. This complexity highlights the challenge of  
 327 accurately distinguish seed abortion using a single gene or variant site.

## 328 **Machine Learning based Genomic Selection for Seedless Grape** 329 **Breeding**

330 Given the polygenic nature of the seedlessness trait, genomic prediction could greatly  
331 improve the speed and accuracy for seedless grape breeding. We extracted the  
332 information of all 794 high-quality variant sites from GWAS analysis, including 77  
333 InDels and 717 SNPs (**Supplementary Table 18**). Using these variations, an unrooted  
334 phylogenetic tree was constructed based on admixed populations ( $n = 444$ ), revealing  
335 that the majority of seedless individuals clustered together (**Fig. 6a**). However, some  
336 seeded samples, like seeded hybrid progeny<sup>20</sup> and Rizamat<sup>44</sup> were mixed in seedless  
337 grapes, as well as seedless samples, such as ‘Dawn Seedless’, ‘Bronx Seedless’,  
338 ‘Cheongsoo’, ‘Ruby Seedless’, and ‘Jingkejing’, were mixed in seeded grapes.  
339 Interestingly, the mutations in the top two QTLs were heterozygous: Chr07: 8.85-8.86  
340 Mb (SDI2 locus) and Chr18: 30.36-31.86 Mb (SDInew locus; **Fig. 6d**). These results  
341 suggest the complexity of seedless in grapes.

342 Therefore, to address this problem, we employed genomic selection based on machine  
343 learning to enhance predictive accuracy (**Extended Data Fig. 14**). We used 794  
344 variant sites and phenotypic data from the admixed populations (444 samples) as the  
345 training dataset, and evaluated the ability of nine different classical models to predict  
346 the seedlessness trait based on 100 rounds of random cross-validations (**Fig. 6b**).  
347 Among these models, machine learning based methods, including SVR-poly and  
348 ElasticNetCV, demonstrated a strong performance in predicting the phenotype,  
349 yielding prediction accuracies of 85.36% and 84.57%, respectively (**Fig. 6b**). As a  
350 result, we applied SVR-poly and ElasticNetCV to genomic prediction on the testing  
351 set data (39 samples not included in the model building) (**Supplementary Table 8**).  
352 We observed that the SVR-poly model and the ElasticNetCV model yielded high  
353 levels of accuracy, with correlation coefficient  $R$  values of 0.99 ( $P < 2.2e-16$ ) and  
354 0.96 ( $P < 2.2e-16$ ), respectively (**Fig. 6c**), suggesting the efficiency of machine  
355 learning based genomic selection of seedless in grapes.



## 356 Discussion

357 Seedless is an important quality trait in table grape breeding. Previous genetic  
 358 investigated the functions of multiple genes, including *VvAGL11*, however,  
 359 quantitative genetic analyses of natural population were not conducted in grapes. In  
 360 this study, we conducted integrative genomic analyses to investigate the polygenic  
 361 basis of seedlessness in grapes: (1) comparative analysis of 15 genomes allowed us to  
 362 discover a heterozygous inversion in Chr10 associated with the seedless trait; (2) the  
 363 evolutionary genomic analyses showed that seedless grapes were closely related with  
 364 an origin from the well-known seedless grape ‘Sultania’ (TS). Introgression rather  
 365 than convergent evolution was associated with the evolution of seedlessness in grapes;  
 366 (3) a total of 110 QTLs associated with 634 candidate genes were identified through  
 367 GWAS analysis within a large natural population, including four significant linkage  
 368 regions such as Chr01: 18.49-19.96 Mb (specific to VV population), Chr07: 8.85-8.86  
 369 Mb (shared, SDI2 locus), Chr18: 11.7-20.0 Mb (specific to VV×VL population), and  
 370 Chr18: 23.9-35.5 Mb (shared, SDInew locus); (4) a total of 339 core genes associated  
 371 with seedlessness was detected through integrative genomics analyses; (5) machine  
 372 learning based genome selection were built to accurately predict the seedless  
 373 phenotypes. Importantly, these findings could efficiently save the cost and time in  
 374 table grape breeding.

### 375 The polygenic nature for seedlessness in grapes

376 The occurrence of seedlessness phenotypes from a cumulative polygenic effect  
 377 associated with different tissues and development stages<sup>17</sup>. The limited observations  
 378 employing single methods, such as transcriptomic analysis, are insufficient. As a  
 379 result, numerous important candidate genes were ignored by previous studies. For  
 380 example, *TT16/ABS*, found at the inversion boundary through comparative genomics  
 381 (**Fig. 1d**), is one of promising candidate genes. Due to the redundant function between  
 382 *STK* and *SHATTERPROOF* (*SHP1* and *SHP2*), the double *abs stk* mutants and triple



383 *tt16 shp1 shp2* mutants all induced fewer seeds and exhibited defects in seed coat  
 384 formation<sup>17,50</sup>. Candidate genes mapped by GWAS, such as *SMC1*, *IIS globulin*  
 385 *G1/2/3*, *LTG1*, *SUS2*, *LAC14*, *TT10/LAC15*, *MANNAN*  
 386 *ENDO-1,4-BETA-MANNOSIDASE 5 (MAN5)*, *PROBABLE FRUCTOKINASE-5*, *E3*  
 387 *UBIQUITIN-PROTEIN LIGASE (DA2)*, *AGL62*, etc., and well-known gene *VviAGL11*,  
 388 play crucial roles in pollen, endosperm and seed coat development (**Fig. 3a, b and**  
 389 **Supplementary Table 11**). Additionally, three transcriptomic analyses, GO  
 390 homologous genes, and previously reported family genes provide us with numerous  
 391 significant candidate genes (**Fig. 5a and Supplementary Table 11**).

392 Importantly, to define the interconnections among multiple genomic analyses,  
 393 integrative genomic analysis was applied in this case, screened out 339 core genes  
 394 with high significance from thousands of candidate genes (**Fig. 5a and**  
 395 **Supplementary Table 17**). Most of these genes are associated with three main tissues  
 396 related to seed coat, endosperm, and embryo, as shown in **Fig. 5b-d**, and ten other  
 397 development processes (**Supplementary Table 11**). For example, we identified  
 398 numerous candidate genes involving in seed hormone regulation, such as *MOTHER*  
 399 *OF FT AND TFL1 (MFT)* in ABA and GA pathways, *VvABI3-1*<sup>68</sup> in ABA pathway,  
 400 *VvGH3.9*<sup>69</sup> in auxin pathways, and *VvMJE1*<sup>64</sup> in jasmonate pathway. Additionally,  
 401 genes controlling the development of floral organs, especially pollen and stigma,  
 402 significantly influence the subsequent ovule development. This includes candidate  
 403 genes like *CYP78A5*, *VvMADS27*, and metal ion transport genes  
 404 *METALLOTHIONEIN-LIKE PROTEINS MT1* and *MT3*<sup>70</sup>. Except for lipid  
 405 accumulation (**Fig. 5d**), the processes of sugar and amino acid synthesis also  
 406 contribute to seed development, such as *BGLU15*, *VviGAPDH*<sup>71</sup>, glycine-rich protein  
 407 (*Vitvi021557*, *Vitvi036421*), and 36.4 kDa proline-rich protein (*Vitvi002605*). Overall,  
 408 these findings not only emphasize the polygenic nature of seedlessness but also  
 409 provide novel candidate genes for functional genetics, highlighting the complexity of  
 410 seed abortion regulation.

## 411 The implications of genomic breeding of seedless table grapes

412 Both individual markers for Marker-Assisted Selection (MAS)<sup>44,72-74</sup> and marker set  
 413 for genomic selection generated in this study could be efficiently used in table grape  
 414 breeding (**Fig. 6d**). Interestingly, in the variation map around *VviAGL11*, we observed  
 415 that all the previously developed molecular markers based on lineages with a narrow  
 416 genetic background for marker-assisted in table grape breeding were filtered away,  
 417 due to either the low genotyping quality or a high missing rate, such as SCF27<sup>75</sup>,  
 418 e7\_VviAGL11<sup>44</sup>, 5U\_VviAGL11<sup>44</sup>, P3\_VviAGL11<sup>18</sup>, VMC7f2<sup>61</sup>, and VrSD10<sup>76</sup> (**Fig.**  
 419 **4d**). Luckily, we designed a set of 12 markers, based on our GWAS analyses of  
 420 natural population with species-wide genetic diversity. These markers could  
 421 accurately delimit seeded and seedless grapes, achieving a precision rate >90% in  
 422 nature populations (**Fig. 4e**).

423 Quantitative genetics analysis revealed 110 high-quality QTLs associated with  
 424 seedlessness in grapevines, including 634 candidate genes (**Supplementary Table 11**).  
 425 Through extracting GWAS significant variants from admixed population, we obtained  
 426 detail information on all 794 significant variant sites for training models. Genome  
 427 selections, employing machine learning algorithms, achieved an impressive precision  
 428 of 99%. This approach could facilitate early genomic selection of natural germplasms  
 429 and hybrid progeny. Many crops, such as tomato<sup>77</sup>, potato<sup>78,79</sup>, cereal<sup>80</sup>, rice<sup>81,82</sup>, wheat  
 430 and maize<sup>83,84</sup>, have successfully applied genomic selection and prediction in their  
 431 breeding programs. However, genomic selection has rarely been used on grape  
 432 breeding. In the future, numerous agronomic traits like fruit aroma, disease resistance,  
 433 soluble solids content, and so on, could be integrated into a single GS chip, offering a  
 434 powerful genomic tool for genomic design of grapevine breeding.

435

## 436 **Methods**

### 437 **Plant materials and genome sequencing**

438 To enrich the genetic diversity of seedless grape varieties, we collected fresh tissues  
 439 of two seedless grape varieties, ‘Thompson Seedless’ (TS) and ‘Black Monukka’  
 440 (BM), from the Anningqu Experimental Station (87°28’00”E, 45°56’00” N) at the  
 441 Xinjiang Academy of Agricultural Sciences in China. Genomic DNA was extracted  
 442 from grape leaves using CTAB method, followed by purification with the QIAGEN  
 443 Genomic kit (CAT#13343). For each sample, a total of 15 µg DNA was utilized for  
 444 HiFi (SMRTbell) library preparation. The sheared DNA fragments (gTUBEs, Covaris,  
 445 USA) underwent treatment with the SMRTbell Enzyme Cleanup Kit (Pacific  
 446 Biosciences, CA, USA) and purification using AMPure PB Beads. The resulting  
 447 libraries were employed for HiFi sequencing on a PacBio Sequel II instrument (CCS  
 448 mode) with Sequencing Primer V2 and Sequel II Binding Kit 2.0 in Grandomics,  
 449 yielding 59.78 Gbp and 60.35 Gbp of sequencing data for TS and BM, respectively.

450 For Hi-C library preparation, the fresh leaves were cut into 2 cm pieces and vacuum  
 451 infiltrated with nuclei isolation buffer supplemented with 2% formaldehyde. The  
 452 isolated nuclei were then digested with 100 units of restriction enzyme DpnII. The  
 453 resulting Hi-C sequencing data amounted to 59.39 Gbp (TS) and 57.67 Gbp (BM) via  
 454 the Illumina Novaseq/MGI-2000 platform. Additionally, the genomic DNA of 29  
 455 grape samples was extracted from fresh leaves, and the high-depth WGS sequencing  
 456 was carried out using the PE150 mode of the Illumina Navaseq 6000 platform. For  
 457 genome annotation, 1 µg total RNA was extracted from mixed tissues, such as roots,  
 458 buds, and leaves. cDNA library was used TruSeq RNA Library Preparation Kit  
 459 (Illumina, USA) and sequenced with 150 bp pair-end reads on the Illumina Navaseq  
 460 6000 platform.

## 461     **Genome assembly**

462     The detailed workflow of haplotype-resolved genome assembly and annotation is  
 463     described in **Extended Data Fig. 1**. The full pipeline for genome assembly and gap  
 464     filling can be found on our lab GitHub@zhouyflab (see Code availability). In brief,  
 465     we utilized the Hi-C and HiFi data integrated assembly algorithm to generate contig  
 466     reads by Hifiasm (v. 0.16.1-r375)<sup>85</sup>. The contig reads were oriented and ordered to  
 467     scaffold level using RagTag (v. 2.1.0)<sup>86</sup> with default parameters. Hi-C reads were also  
 468     employed to anchor scaffolds onto chromosomes by Juicer 2.0<sup>87</sup> and Juicebox (v.  
 469     1.11.08)<sup>88</sup>. The adjusted genome at the chromosome level was generated using  
 470     3D-DNA (v. 201008)<sup>89</sup>. The detailed workflow of haplotype-resolved genome  
 471     assembly and annotation is described in **Extended Data Fig. 1**. The full pipeline for  
 472     genome assembly and gap filling can be found on our lab GitHub@zhouyflab (see  
 473     Code availability). The chloroplast genome was de novo assembly using GetOrganelle  
 474     toolkit (v1.7.7.0)<sup>90</sup> with K-mer parameters set to 21, 65, 105, and 127. For the  
 475     high-quality mitochondrial genome, de novo assembly was performed using Flye  
 476     (v2.9.2)<sup>91</sup> with HiFi long reads. Furthermore, using PN\_T2T genome<sup>46</sup>, we utilized  
 477     RagTag to assemble three chromosomes-level genomes based on scaffold level:  
 478     ‘Cabernet Sauvignon’ (CS)<sup>52</sup>, ‘Black Corinth Seedless’ (BC seedless)<sup>92</sup>, and ‘Black  
 479     Corinth Seeded’ (BC seeded)<sup>92</sup>.

## 480     **Assembly assessment**

481     HiFi data from TS and BM were performed to assess genome heterozygosity based on  
 482     k-mer using GenomeScope2.0<sup>93</sup>. Meanwhile, basic genome statistics were calculated  
 483     using seqkit (v. 2.2.0)<sup>94</sup>, which included genome length, N50, and GC content. The  
 484     completeness of the haplotype genomes was evaluated using BUSCOs (v. 5.3.0)<sup>95</sup>  
 485     with the embryophyta\_odb10 database. Merquy (v. 1.3, best\_k=19)<sup>96</sup> was employed  
 486     to evaluate the quality value (QV) and completeness of the haplotype genomes based  
 487     on whole-genome sequencing data. The Hi-C interactive signals on grape genome

488 were visualized using Juicebox.

## 489 **Genome annotation**

490 The genome-wide annotation pipeline, identification of telomeres and centromeres,  
491 and annotation of transposable elements (TEs) were referred from pervious study<sup>46,97</sup>.  
492 The statistic results of TEs classification through Pan-genome TE annotation<sup>46</sup> can be  
493 found in **Supplementary Table 2**. Additional details on telomere regions, telomere  
494 copy numbers, and centromere regions for each chromosome are provided in  
495 **Supplementary Table 3**. The CS genome was annotated based on sequence similarity  
496 using Liftoff (v. 1.6.3)<sup>98</sup> and PN\_T2T annotation files.

## 497 **Comparative genomics**

498 For genome level variant calling, we selected a total of 15 grape genomes, including  
499 ten seeded accessions: ‘PN40024’ (PN\_T2T), ‘Cabernet Sauvignon’ (CS, hap1 and  
500 hap2)<sup>92</sup>, ‘Muscat Hamburg’ (MH, hap1 and hap2), ‘Shine Muscat’ (SM, hap1 and  
501 hap2), ‘Muscadinia rotundifolia’ (MR, hap1 and hap2)<sup>99</sup>, BC Seeded<sup>92</sup>, as well as five  
502 seedless: TS (hap1 and hap2), BM (hap1 and hap2), BC Seedless<sup>92</sup>. Among them, TS  
503 and BM were newly sequenced in this study. Four recently synchronized  
504 haplotype-resolved genomes of MH and SM will be published in another study.

505 All chromosome-level genomes were aligned with PN\_T2T genome using Mummer4  
506 (v. 4.0.0rc1)<sup>100</sup>, and the results were visualized using Plotsr (v. 0.5.4)<sup>101</sup> and Linux  
507 based Gunplot. R script was used for visualizing the gene density (line) of the  
508 reference genome (see Code availability). To validate the authenticity of SVs, the raw  
509 reads of HiFi and Hi-C were mapped to their haplotype genomes using Minimap2.  
510 The corresponding BAM files were extracted using SAMtools (v. 1.13)<sup>102</sup> and then  
511 inputted into the IGV software for inversions validation. In addition, the genome  
512 consensus phylogenetic tree was constructed using OrthoFinder (v. 2.5.4)<sup>103</sup> based on  
513 the single-copy orthologous genes from the whole genomes.

## 514 **WGS variation detection**

515 To genotype SNPs, InDels and SVs in 548 accessions, low-quality resequencing reads  
 516 were removed using Fastp (v. 0.23.2)<sup>104</sup> with default parameters. The filtered data  
 517 were then mapped to the CS reference genome using BWA (v. 0.7.17-r1188)<sup>105</sup>.  
 518 Non-uniquely mapped and duplicated reads were excluded using SAMtools and  
 519 GATK (v.4.2.3.0)<sup>106</sup>. Subsequently, SNPs and InDels calling were performed using  
 520 GTX (v. 2.1.11, <http://www.gtxlab.com/product/cat>), followed by the merging of  
 521 genotyping of the gVCF files into a VCF file. Delly (v. 1.1.6)<sup>107</sup> was used to SVs  
 522 calling with default parameters. Basic filtering of the VCF file was performed using  
 523 VCFtools (v. 0.1.16)<sup>108</sup> with the following parameters: --max-missing 0.8, --minGQ  
 524 20, --min-alleles 2, --max-alleles 2, --minDP 4, --maxDP 1000, and --maf 0.0005.  
 525 PLINK (v. 1.90b6.21)<sup>109</sup> was utilized to reduce the size of the VCF file and improve  
 526 computational efficiency. Finally, we obtained a total of 4,462,797 SNPs, 443,812  
 527 InDels, and 487,204 SVs from 548 grape accessions in the nuclear genome, 38,267  
 528 variation sites (SNPs and InDels) from 314 grape accessions in the MT genome, and  
 529 2,247 variation sites (SNPs and InDels) from 314 grape accessions in the Pltd  
 530 genome.

## 531 **Population genetics analysis**

532 Phylogenetic analysis was conducted using 250,821 high-quality SNPs filtered for LD  
 533 by PLINK, with parameters: --indep-pairwise 20 5 0.2 --geno 0.1. The phylogenetic  
 534 tree was inferred by Iqtree (v. 2.1.4-beta)<sup>110</sup> with parameters: -m GTR+I+G -bb 1000  
 535 -bnni -alrt 1000 -st DNA, and the result were visualized using iTOLs<sup>111</sup>. Similarly, the  
 536 phylogenetic analysis of the MT and PT genomes yielded 9,647 and 758 high-quality  
 537 SNPs using PLINK (--geno 0.2 --maf 0.001), respectively. The PCA and IBD analysis  
 538 were preformed using PLINK and visualized using R scripts and Cytoscape (v.  
 539 3.9.1)<sup>112</sup>, respectively. VCFtools was employed to calculate the  $F_{ST}$  and  $\pi$  statistics at  
 540 whole genome level with 20 kb window size. Population introgression analysis was

541 calculated using the "ABBABABAwindows.py" script from the "general\_genomics"  
542 tool ([https://github.com/simonhmartin/genomics\\_general](https://github.com/simonhmartin/genomics_general)) with 20 kb window size.  
543 The population branch statistic (PBS) assesses differentiation in the same branch of  
544 the phylogenetic tree with 50 SNP per window between seeded and seedless grapes in  
545 both VV and VV×VL populations<sup>113</sup>. PBSscan was utilized to estimate population  
546 differentiation using  $D_{xy}$  (-div 2).

#### 547 **Genome-wide association study**

548 We selected three populations for GWAS analysis, including VV population (n = 352),  
549 VV×VL population (n = 92), and the admixed population (n = 444). High-quality  
550 variants were obtained using PLINK with MAF ≥ 0.05 and missing < 0.2, resulting in  
551 2,086,600 variants (1,881,457 SNPs, 173,547 InDels, and 31,596 SVs), 1,419,982  
552 variants (1,274,130 SNPs, 124,054 InDels, and 21,798 SVs), and 2,292,404 variants  
553 (2,065,307 SNPs, 192,536 InDels, and 34,561 SVs), respectively. GWAS analysis  
554 utilized the mixed linear model in GEMMA (v. 0.98.3)<sup>114</sup> with the first three PCAs as  
555 a random effect matrix. Variant positions and p-wald test statistics (*P*-value) were  
556 extracted to generate Manhattan plots and Q-Q plots by R scripts. A Python script was  
557 utilized to identify the associated genes within 5 kb windows of the significant  
558 variants, which were above the significance threshold of  $-\log_{10}(0.05/\text{Variant}$   
559  $\text{Numbers})$ .

560 GO enrichment analysis was employed using the web tool DAVID<sup>115</sup>, and the results  
561 were visualized with R script. The SDI locus was identified through molecular  
562 markers SNP-25.24 and SNP-26.93<sup>20</sup>. The LD linkage heatmap was visualized using  
563 LDBlockshow (v. 1.40)<sup>116</sup>. Moreover, DIAMOND (v. 2.0.15)<sup>117</sup> was employed for  
564 sequence comparisons of whole-genome homologous protein from 14 genomes  
565 (Seeded: Seedless: Outgroup = 8:5:1). Sequence alignments of candidate proteins  
566 were conducted using ClustalW in MEGA11<sup>118</sup>, and sequence visualization was  
567 accomplished using GeneDoc (v. gd322700)<sup>119</sup>. The visualization of FPKM values,

gene structures, and mutation ratios was achieved using R scripts (see Code availability).

### **Multi-transcriptome analysis**

Three independent transcriptome datasets were downloaded from the NCBI database (see **Supplementary Table 15**). The 76 samples encompassed six grape varieties: PN and TS, which underwent repeated testing from 20 DAF to 50 DAF over a two-year period; 'Italia' and HS, continuously evaluated from 7 to 42 DAF within a single year; and Himrod (VV×VL population) and Jinzao (VV population), continuously monitored from full flowering stage to grape maturity within a single year. These raw fastq data underwent quality control and data cleaning using Fastp with default parameters. Using the PN\_T2T genome as a reference, transcriptomic assembly was conducted on the processed data using STAR (v. 1.5.2)<sup>120</sup>. The R package DESeq2 (v.1.36.0)<sup>121</sup> was utilized for PAC analysis and pairwise comparisons between seeded and seedless grape samples at each development time point. The thresholds for differential expression genes were  $|\text{Log2FoldChange}| \geq 2$  and  $P\text{-adjust} < 0.05$ .

### **Integrative genomic analysis**

Homologous protein alignment identified 14,650 genes within the 34 seed development-related GO terms (EMBL-EBI, QuickGO database, 2023-01-03), and 6,529 genes with high sequence similarity were screened (identity  $\geq 50\%$ , e-value  $\leq 1e-5$ ). Additionally, the primer sequences collected from previous studies in the past decade, including 451 genes and 7 molecular markers associated with seed abortion processes. All of them were then mapped to the PN\_T2T genome based on sequence alignment using BALST in TBtools (v. 1.113)<sup>122</sup>. The results were visualized using Venn diagrams and whole-genome density plots using the R package ggvenn (v0.1.10) and RIdeogram<sup>123</sup>, respectively. The integrated results in this study, including GWAS results, RNA-seq results, GO enrichment results, and previously reported family genes, were overlapped and filtered manually. The core candidate genes were



595 screened based on an average expression value across all time points (AVE FPKM  $\geq$   
596 100) and Fold-Changes (Fold = Seeded AVE FPKM / Seedless AVE FPKM, Fold  $\leq$   
597 0.5 or Fold  $\geq$  2.0). Data visualization was performed using an R script, which  
598 included UpSet plot analysis and gene expression heatmaps (see Code availability).

## 599 **Genome selection based on machine learning**

600 For genome selection, we utilized the admixed population (n = 444) as training set,  
601 and additional 39 samples as testing set for phenotype prediction, as pipeline showed  
602 in **Extended Data Fig. 14**. Beagle (v. 5.2\_21Apr21.304.jar)<sup>124</sup> was used to impute the  
603 VCF file. Subsequently, the 794 variants, above the significance threshold of  
604  $-\log_{10}(0.05/\text{Variant Numbers})$  ( $\sim 7.66$ ), were extracted from imputed VCF using  
605 BCFtools (v. 1.13)<sup>125</sup>, including 717 SNPs and 77 InDels. For model selection, we  
606 utilized the Python package ‘sklearn’ (<https://scikit-learn.org/stable/install.html>) and  
607 selected nine classical models: Cross-validated Elastic Net model (ElasticNetCV),  
608 Kernel Ridge Regression (KernelRidge), Lasso Regression (Lasso), Linear  
609 Regression (Linear), Logistic Regression (Logistic), PLS Regression, Linear Ridge  
610 Regression (Ridge), Linear Support Vector Regression (SVR-Linear), and Polynomial  
611 Support Vector Regression (SVR-poly). After 100 rounds of random cross-validation  
612 with the training set, we chose the models with best performance, SVR-poly and  
613 ElasticNetCV, for the prediction of testing set. Moreover, the Iqtree was employed to  
614 construct an unrooted tree, while iTOLs was used for the visualization of the  
615 phylogenetic tree. The data visualization used R scripts.

## 616 **Figure Legends**

617 **Fig. 1 | Comparative genomic between seeded and seedless grape cultivars. a,**  
618 Visualization of the haplotype-resolved genomes aligned with the PN\_T2T, with  
619 discernible inversions marked in yellow. **b** and **c**, Morphological features of TS and  
620 BM. **d**, Comparative genomic analysis of 15 grape genomes prioritized by the  
621 phylogenetic tree constructed with single-copy gene features. Included genomes: MR

622 (Muscadinia rotundifolia), CS (Cabernet Sauvignon), MH (Muscat Hamburg), SH  
623 (Shine Muscat), PN (PN40024), BC (Black Corinth). **e**, Detailed validation of a  
624 prominent Chr10 segment inversion, with its distinctive features illustrated in the  
625 Hi-C heatmap. **f**, Gene loss in hap2 genomes within the Chr10 inversion region.  
626 *VvSUS2* and *VvTT16* are located near the inversion boundaries. Red blocks represent  
627 lost genes, while gray blocks indicate genes shared between the two haplotype  
628 genomes.

629 **Fig. 2 | Evolutionary genomics of seedlessness in grapevine.** **a**, Comparison of  
630 grape clusters and cross-sectional views of berry from three cultivated grape varieties:  
631 TS, BM, and CS. The scale bar indicates 5 cm for grape clusters and 1 cm for  
632 individual berries. **b**, PCA analysis on the whole-genome sequencing (WGS) data,  
633 except the outgroup, resulting in obviously separated five populations. Solid dots  
634 denote seedless grapes, while transparent dots represent seeded grapes. CS genome  
635 serves as the reference genome for variants calling. **c**, Phylogenetic tree analysis of  
636 the six populations. The light-blue blocks represent seedless grapes and star symbols  
637 indicate TS and BM (see Extended Data Fig. 6 for detailed information of the  
638 phylogenetic tree). **d**, IBD analysis of 46 seedless grapes, filtering the results with a  
639 threshold of 0.40. Purple points represent the *V. vinifera* × *V. labrusca* population,  
640 while red points represent the *V. vinifera* population. **e**, The phylogenetic tree of 46  
641 seedless grapes.

642 **Fig. 3 | GWAS mapping of the polygenic basis of seedlessness.** **a** and **b**,  
643 Seedless-associated genomic loci and genes across three populations: the VV  
644 population (in red, n=352), the VL population (in purple, n=92), and the admixed  
645 population (in yellow and dark-blue, n = 444). The horizontal dashed lines denote the  
646 Bonferroni thresholds ( $-\log_{10}[0.05/\text{Variant Numbers}]$ ): 7.62 for VV, 7.45 for VV × VL,  
647 and 7.66 for admixed population, respectively. Points represent SNPs, while triangles  
648 represent SVs (and InDels). **c**, LD correlation analysis of a highly linked SDI2 locus  
649 in Chr07, and genes associated with seed abortion are highlighted in red. **d**, Admixed

650 population analysis of a highly linked region in Chr18. The SDI locus (Chr18:  
651 29.83-31.34 Mb) is identified based on SNP markers, and the redefined SDI locus  
652 (SDInew, Chr18: 30.36-31.86 Mb) is defined based on GWAS results, depicted in the  
653 grey block. The significant threshold: 7.61 for SNPs, and 6.66 for SVs (and InDels).  
654 The dashed line for fixation indices ( $F_{ST}$ ) indicates an average value of 0.126, with  
655 genetic diversity ( $\pi$ ) measured 0.0019 for VV×VL and 0.0016 for VV. The top 5th  
656 percentile of  $f_d$  statistics is 0.277, with PBS statistics measuring 0.356 in VV and  
657 0.415 in VV×VL. **e**, Gene flow pattern based on ABBA-BABA statistics, where SD  
658 represents Seeded, SL represents Seedless.

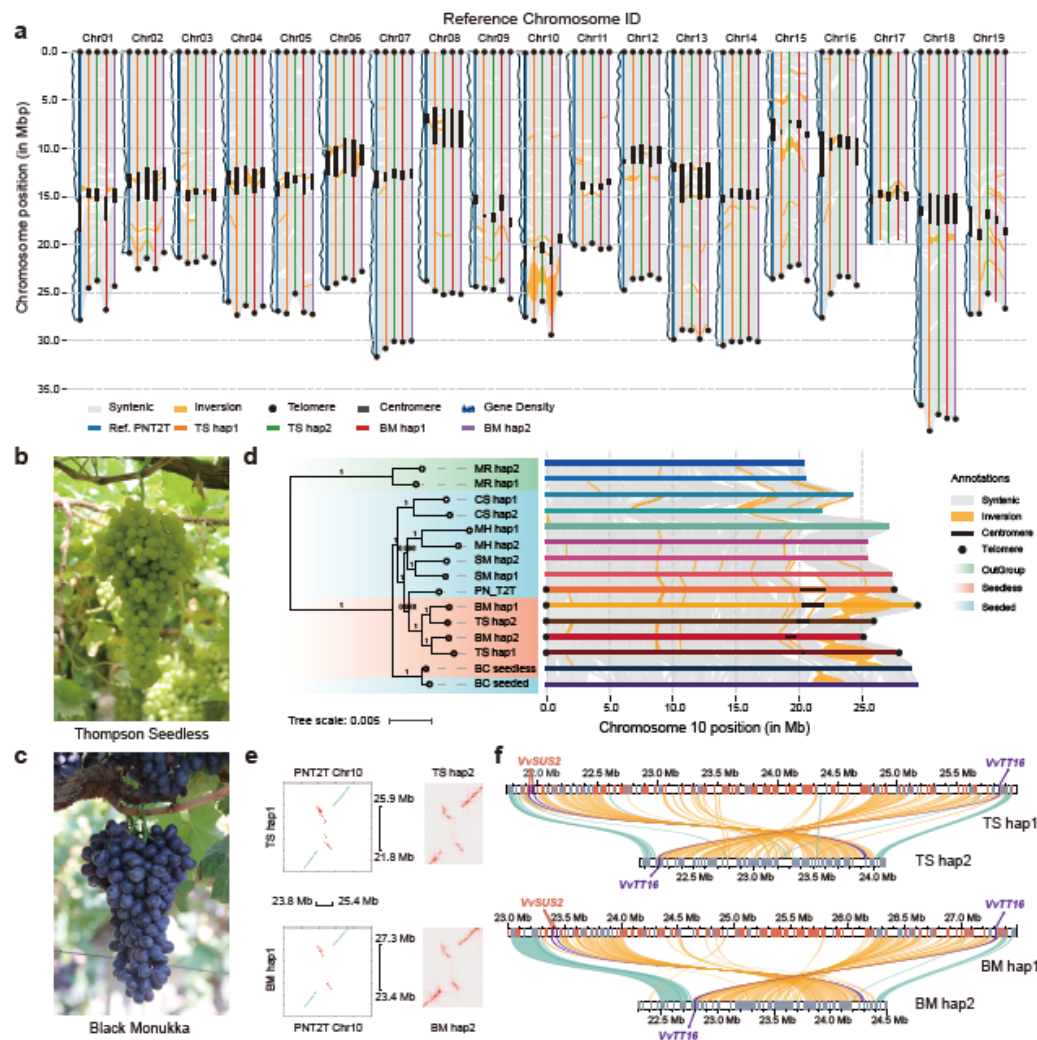
659 **Fig. 4 | Deep mining of key loci associated with seed abortion.** **a**, Sequence  
660 alignment of three *IIS GLOBULIN SEED STORAGE PROTEIN* genes in Chr07  
661 across 14 grape varieties. Black blocks represent deletion and nonsynonymous  
662 mutations. **b**, Relative expression values of these three genes at different time points  
663 across six grape varieties. Abbreviations: DAF (Day after flowering), FF (Full  
664 flowering), BE (Berry expansion), V (Veraison), M (Maturity). **c** and **d**, Visualization  
665 of crucial candidate genes associated with seed abortion within the SDInew locus  
666 region. GWAS significance thresholds ( $-\log_{10}[0.05/\text{Variant Numbers}]$ ): 7.61 for SNPs,  
667 and 6.66 for SVs (and InDels). The top ten percentiles of significant variants are  
668 Chr18\_31295826 ( $y = 26.39$ ). **e**, Genotyping percentage of highly significant variants  
669 is within the region Chr18: 30.70-31.32 Mb. ‘Seedless’ includes cases with genotypes  
670 0/1 and 1/1, ‘Seeded’ includes the case with genotype 0/0.

671 **Fig. 5 | Integrative genomic analyses for grapevine seed abortion.** **a**, Results from  
672 integrative genomic analyses: GWAS, transcriptomic analysis, reported genes and  
673 markers mapping, and GO homologous genes overlapping. Blue bars represent the  
674 number of genes uniquely and overlappedly through this approach. **b** and **c**, Heatmap  
675 of  $\text{Log}_2(\text{FPKM})$  for candidate genes related to embryo, endosperm, and seed coat  
676 development, resulting from integrative genomic analyses. The genes are indicated in  
677 relation to their expression in specific tissues and time points on schematic

678 representation transverse profiles of seeds. **d**, Expression patterns of genes involved in  
 679 lipid synthesis, degradation and transportation, which also pinpointed on the  
 680 schematic representation of oil body formation. Abbreviations: DAF (Day after  
 681 flowering), FF (Full flowering), BE (Berry expansion), V (Veraison), M (Maturity).

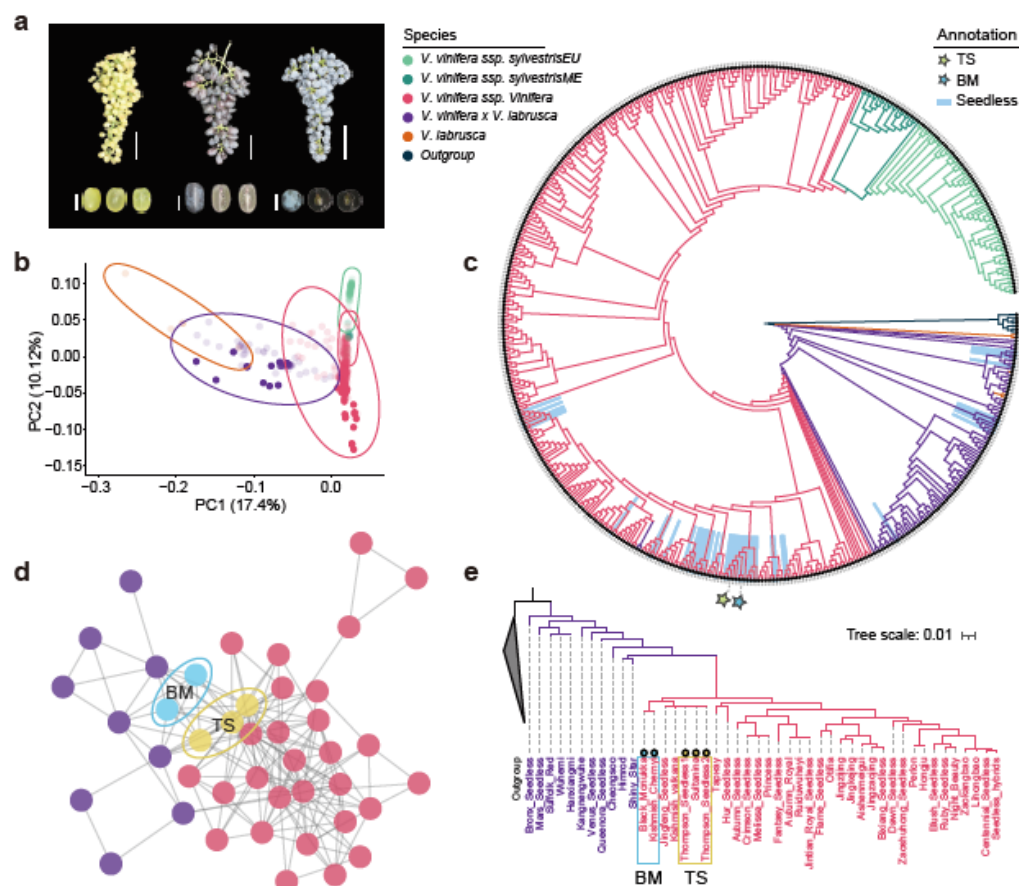
682 **Fig. 6 | Machine learning based genomic selection on seedlessness in grapevine**  
 683 **breeding. a**, Phylogenetic clustering based on 794 significant variants (77 InDels and  
 684 717 SNPs) derived from the GWAS analysis. **b**, Comparison of seedlessness  
 685 prediction accuracy across nine classical models for genome selection. The training  
 686 set comprises of 444 grape samples and their phenotypes. **c**, Prediction results of the  
 687 two best-performing models. The testing set includes 39 samples, distinct from the  
 688 444 samples used for model training. **d**, Genotyping visualization of the 794 variants,  
 689 sorted based on prediction results of the SVR-ploy model.

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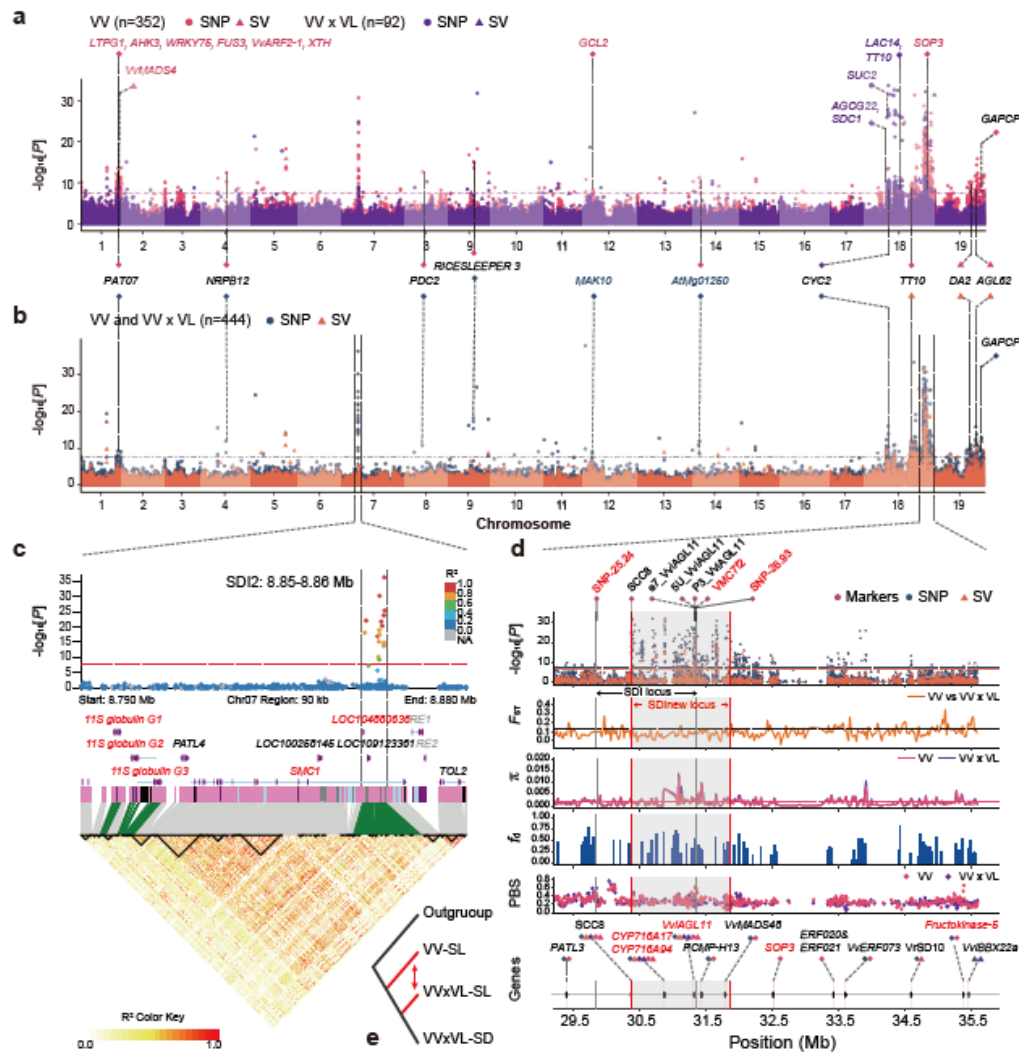
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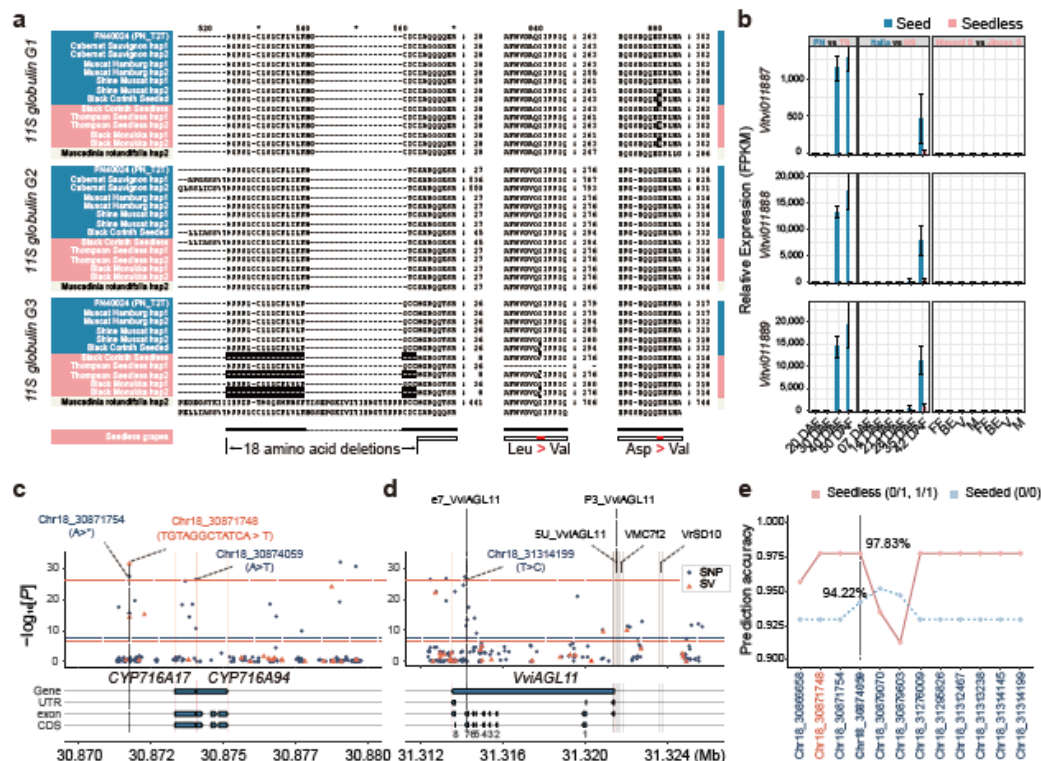


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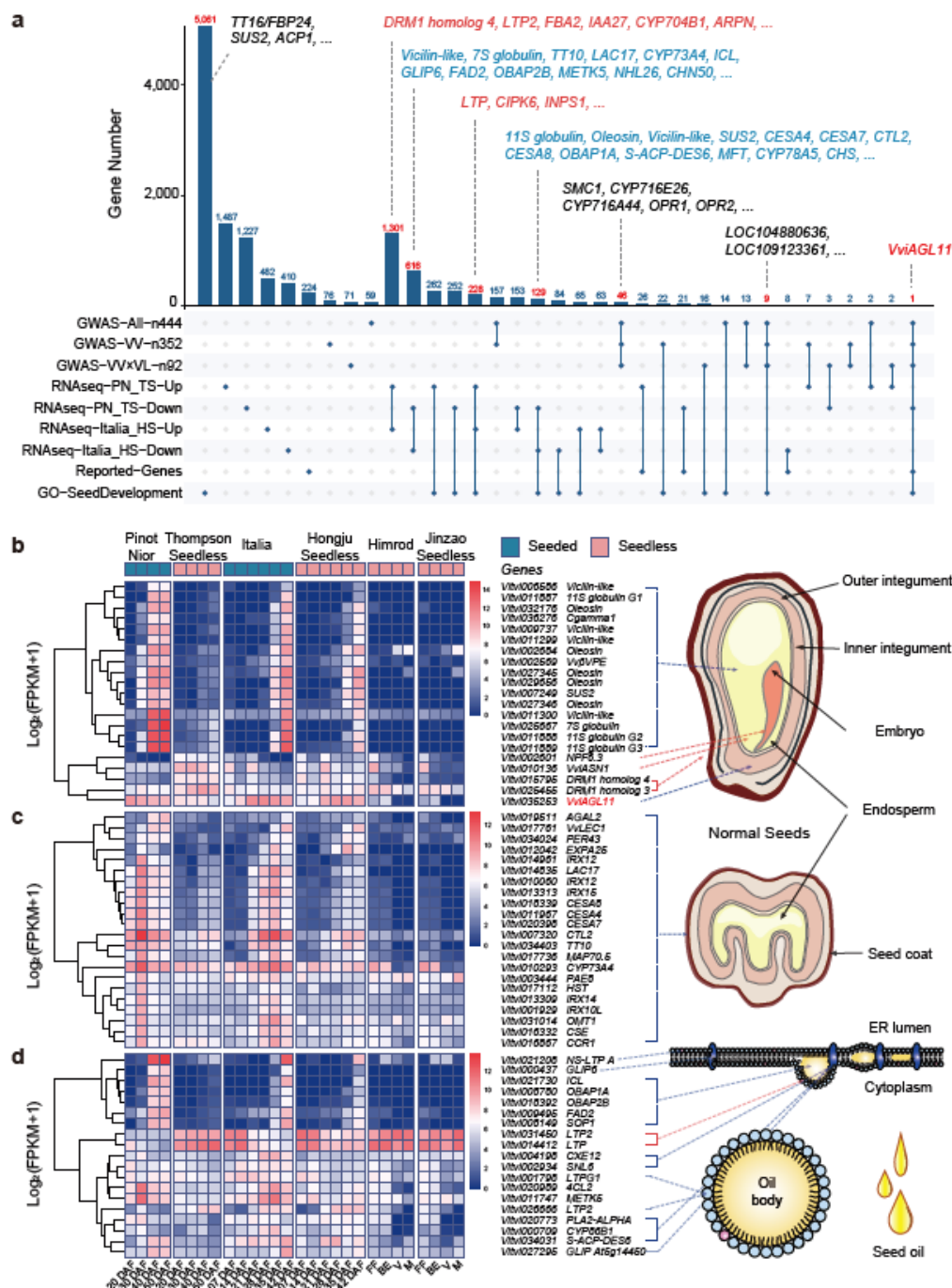


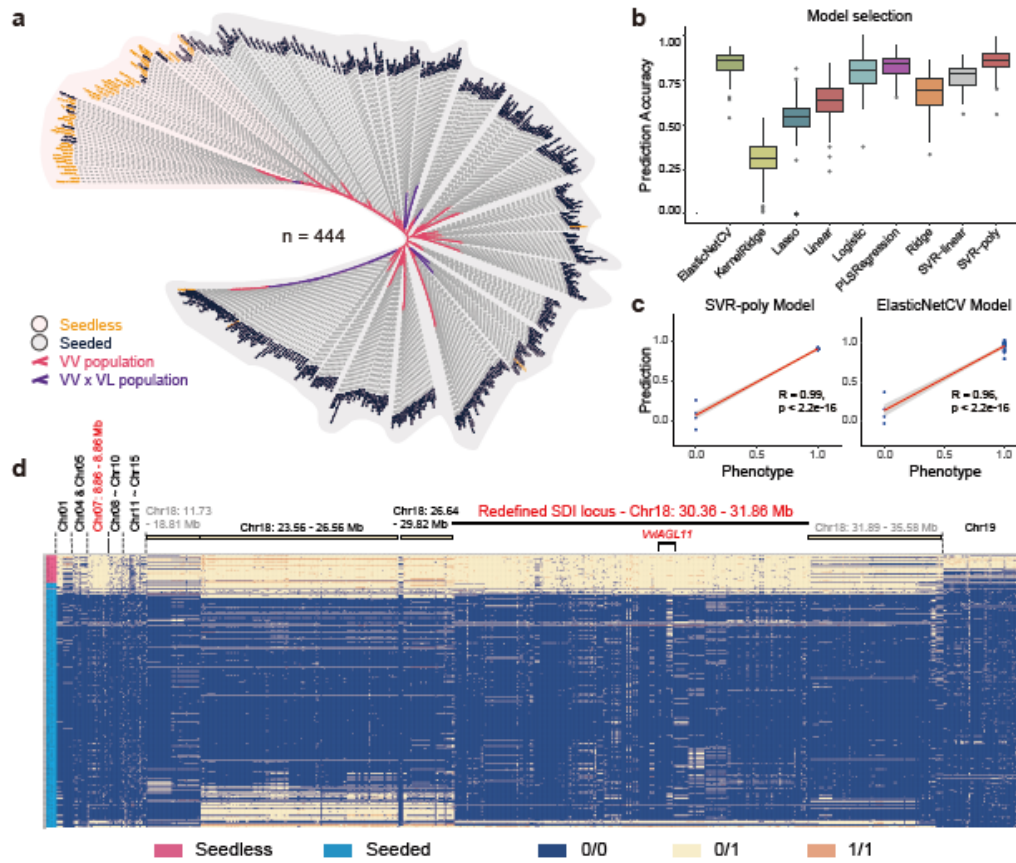


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## 703    **Supplementary Data**

704    **Extended Data Fig. 1 | Complete workflow for haplotype-resolved genome**  
 705    **assembly and annotation.** Additional details can be found on our lab website  
 706    GitHub@zhouyflab.

707    **Extended Data Fig. 2 | Evaluation of four haplotype-resolved genomes. a,**  
 708    BUSCO assessment of genome completeness using the embryophyta\_odb10 database.  
 709    **b,** Evaluation of quality value (QV) and haplotype completeness using Merqury based  
 710    on k-mer. **c,** Visualization of the diploid genome via Hi-C heatmap using Juicebox.  
 711    Most single contig (green rectangle) are directly composed of chromosomes (blue  
 712    rectangles).

713    **Extended Data Fig. 3 | Comparative genomics results. a,** Sequence alignment of 15  
 714    grape genomes with the PN\_T2T genome. Red blocks represent seedless samples,  
 715    green blocks indicate seeded samples, and gray blocks denote outgroup samples. Red  
 716    stars highlight inversions associated with seed abortion. The Chr07 of '*Muscadinia*  
 717    *rotundifolia*' is composed of Chr07 and Chr20. **b,** Sequence alignment results of  
 718    Chr15 for the 15 genomes, as well as the inversion Hi-C heatmap in inversion  
 719    boundary. The phylogenetic tree was constructed using single-copy genes from the  
 720    whole genome proteins.

721    **Extended Data Fig. 4 | Reads mapping at the inversion breakpoints in seedless**  
 722    **haplotype genome. a, c, and e,** represent the start points of the inversions, while **b, d,**  
 723    and **f,** represent the end points of the inversions. Coverage depth is halved before and  
 724    after the breakpoint junctions, revealing transitions between heterozygous and  
 725    homozygous states of reads sequences are observed.

726    **Extended Data Fig. 5 | Detailed phylogenetic tree of 548 grapevine accessions.**  
 727    This figure provides a full zoomed-in version of **Fig. 3c**, which includes the six  
 728    populations. Light-blue blocks represent seed abortion samples and black star

729 symbols indicate TS and BM.

730 **Extended Data Fig. 6 | Phylogenetic tree of mitochondrial and chloroplast**  
 731 **genomes in 314 grapevine accessions.** a-c, represent the consensus phylogenies  
 732 constructed based on the mitochondrial genomes, nuclear genomes, and chloroplast  
 733 genomes, respectively. d-e, depict the complete phylogenetic trees of the  
 734 mitochondrial and chloroplast, encompassing six populations.

735 **Extended Data Fig. 7 | Visualization of whole-genome analyses.** From top to  
 736 bottom, these results included QTL peaks (red: VV, purple: VV×VL, black:  
 737 consensus peaks), GWAS analyses within three populations (red: VV, purple: VV×  
 738 VL, dark-blue and yellow: admixed population; points: SNPs, triangles: InDels and  
 739 SVs), population analyses of fixation indices ( $F_{ST}$ ), nucleotide diversity ( $\pi$ ),  
 740 introgression ( $f_d$ ), and divergent selection (PBS) (refer to Supplementary Table 13),  
 741 and 339 core candidate genes (refer to Supplementary Table 11).

742 **Extended Data Fig. 8 | Quantile-Quantile (Q-Q) plot and Gene Ontology (GO)**  
 743 **enrichment analysis.** Genome-Wide Association Study (GWAS) Q-Q plot for the  
 744 three populations, along with GO enrichment analysis for biological processes in  
 745 genes specific to VV×VL and VV, as well as those consensus gene for admixed  
 746 populations (refer to Supplementary Table 12).

747 **Extended Data Fig. 9 | Sequence alignment of *11S globulin G1-G3* homologous**  
 748 **genes from 14 grape genomes:** includes ‘PN\_T2T’ (PN40024), ‘Cabernet Sauvignon’  
 749 (CS), ‘Muscat Hamburg’ (MH), ‘Shine Muscat’ (SM), ‘Black Corinth Seeded’  
 750 (BCsd), ‘Black Corinth Seedless’ (BCsl), ‘Thompson Seedless’ (TS), ‘Black  
 751 Monukka’ (BM), and ‘Muscadinia Rotundifolia’.

752 **Extended Data Fig. 10 | Three candidate genes related to seedlessness in Chr18.**  
 753 a-c, Candidate genes were identified through GWAS analysis using the admixed  
 754 population, with a significant threshold of 7.61 for SNPs and 6.66 for SVs (and

755 InDels).

756 **Extended Data Fig. 11 | Principal Component Analysis (PCA) analysis of the**  
 757 **three transcriptomic datasets, and relative expression values (FPKM) for**  
 758 ***VviAGL11*.**

759 **Extended Data Fig. 12 | Enrichment analysis of GO homologous genes related to**  
 760 **grape seed development. a**, Enrichment analysis of 14,650 seed development genes  
 761 from the GO database. **b**, Enrichment analysis of 6,529 GO homologous genes  
 762 associated with grape seed development.

763 **Extended Data Fig. 13 | Visualization of multiple seed development-related**  
 764 **datasets. a** and **b**, Results from two independent transcriptomic analyses, ‘Italia’ vs  
 765 ‘Hongju Seedless’ (HS) and ‘Pinot Noir’ (PN) vs ‘Thompson Seedless’ (TS),  
 766 overlapping with GO homologous gene. Red indicates up-regulated DEGs, blue  
 767 represents down-regulated DEGs, and yellow denotes GO homologous genes  
 768 associated with seedlessness. **c**, Results from integrative genomic analyses: GWAS,  
 769 transcriptomics, reported genes mapping, and GO homologous genes. Green bars  
 770 represent the total number of genes identified through this approach. **d**, Visualization  
 771 of three datasets: GO homologous genes, reported gene families, and reported  
 772 molecular markers.

773 **Extended Data Fig. 14 | Genome selection workflow.** The 794 significant variants  
 774 extracted from GWAS results, including 77 InDels and 717 SNPs. More detail code  
 775 can be found on our lab GitHub@zhouyflab.

776

777 **Supplementary Table 1. Assessment of genome quality.** A comparison between  
 778 four T2T haplotype-resolved genomes and the completed reference genome PN\_T2T.

779 **Supplementary Table 2. Pan-genome TE annotation results in the four**  
 780 **haplotype-resolved genomes.**

781 **Supplementary Table 3. Centromere and telomere regions in the four**  
 782 **haplotype-resolved genomes.** This information includes the start position, end  
 783 position, copy numbers, and TRF ID.

784 **Supplementary Table 4. Comparative genomic statistics.** Genome alignment was  
 785 conducted using the SyRI (v. 1.5.4) with input from the output file generated by  
 786 Mummer4 (v. 4.0.0rc1).

787 **Supplementary Table 5. Summary of genome annotation in the TS hap2 genome**  
 788 **region Chr15: 8.72-9.90 Mb.** This region contains a total of 111 genes, including 73  
 789 shared genes and 38 genes that are gained in the TS hap1 genome.

790 **Supplementary Table 6. Summary of genome annotation in the TS hap1 genome**  
 791 **region Chr10: 21.75-26.00 Mb.** This region contains a total of 210 genes, including  
 792 79 shared genes and 131 genes that are exclusively lost in the TS hap2 genome.

793 **Supplementary Table 7. Summary of genome annotation in the BM hap1**  
 794 **genome region Chr10: 23.00-27.50 Mb.** This region contains a total of 237 genes,  
 795 including 69 shared genes and 168 genes that are exclusively lost in the BM hap2  
 796 genome.

797 **Supplementary Table 8. Population information for grape resequencing.** This  
 798 table provides details, such as NCBI accessions, the full name of sample, species,  
 799 seed conditions, population used in GWAS, samples used in phylogenetic trees,  
 800 samples for population analyses ( $F_{ST}$ ,  $\pi$ ,  $f_d$ , and PBS), the training set and testing set  
 801 for genome selection, as well as prediction results of best-performance models.

802 **Supplementary Table 9. Identity by Descent (IBD) matrix for 46 seedless**  
 803 **individuals.** In this matrix, the VV population is denoted by purple, while VV×VL  
 804 population is indicated by red. Detailed information for all samples used can be found  
 805 in Supplementary Table 8.

806 **Supplementary Table 10. GWAS results of three populations.** This table includes  
807 the variant positions, allele changes, -Log10 (*P* value), and associated genes (within ±  
808 5 kb of the variant sites).

809 **Supplementary Table 11. Integrative genomic analysis across all study results.**  
810 This table consolidates GWAS analysis results in different populations, differentially  
811 expressed genes from three transcriptomic analyses, grape seed development  
812 associated GO homologous genes, reported gene families and molecular markers, as  
813 well as 339 core candidate genes identified through integrative genomic analysis. The  
814 reference genome utilized ‘Cabernet Sauvignon’ (CS), and the homologous proteins  
815 was aligned with the UniProt database and PN40024 12X (GCF\_000003745.3),  
816 respectively.

817 **Supplementary Table 12. Distinct and consensus genes, and GO enrichment**  
818 **analyses in three populations.** This table presents candidate genes specific to the VV  
819 population, those specific to VV×VL population, and those consensus in the admixed  
820 population as identified through GWAS. GO enrichment analysis was conducted  
821 using the online toolkit DAVID (<https://david.ncifcrf.gov/tools.jsp>).

822 **Supplementary Table 13. Population analyses results.** This table includes the  
823 genome-wide fixation indices ( $F_{ST}$ ) analysis for 11 VV and 35 VV×VL seedless  
824 samples, genetic diversity ( $\pi$ ) analysis for two population (11 VV and 35 VV×VL  
825 seedless samples),  $f_d$  statistics analysis for gene introgression, and population branch  
826 statistic (PBS) analysis for VV and VV×VL populations. Samples were selected from  
827 a branch in the phylogenetic tree, with wild grapes (ME) as the outgroup. The  
828 statistical window size is 20 kb for  $F_{ST}$ ,  $\pi$  and  $f_d$ , while 50 SNPs per window size for  
829 PBS analysis. Population information can be found in Supplementary Table 8.

830 **Supplementary Table 14. Mutation ratio statistics of significant variants**  
831 **identified from GWAS analysis within Chr18:30.70-31.32 Mb in the admixed**  
832 **population.**



833 **Supplementary Table 15. Information on three transcriptomic groups, including**  
834 **NCBI accessions, full name of samples, time points, species and so on.**

835 **Supplementary Table 16. Primer sequences and data sources for 451 family**  
836 **genes and 7 molecular markers, mapping proteins with the PN\_T2T genome.**

837 **Supplementary Table 17. Overlapped genes statistics used for upset plot.** Further  
838 gene details can be found in Supplementary Table 11.

839 **Supplementary Table 18. Genotyping information of 794 high-quality variants**  
840 **(77 InDels and 717 SNPs).** This table also includes prediction results using genome  
841 selection based on the SVR-poly model. ‘0’ represents 0/0, ‘1’ represents 0/1, and ‘2’  
842 represents 1/1.

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849 assembly.

## 850 **Author contributions**

851 Y. Z. conceived and designed the study. H. Z., F. Z. and X. W. collected the plant  
852 materials. H. Z. and Z. L. performed experiments and genome sequencing. X. W., S.  
853 C., Y. S., T. H., and Y. Z assembled and annotated the haplotype-resolved T2T  
854 genomes. X. W., Z. L., F. Z., X. H., W. L., Z. L., and Y. G. performed data analysis. Y.  
855 Z., X. W., Z. L., F. Z., H. X. wrote the original draft of the manuscript. All authors  
856 provided critical feedback and revised the manuscript. X. W. and Z. L. contributed  
857 equally to this work.



## 858 **Competing interests**

859 The authors declare no competing interests.

## 860 **Data availability**

861 The raw sequencing data, comprising PacBio HiFi long-reads, Illumina Hi-C reads,  
862 RNA-seq reads, and 29 WGS grape accessions, is accessible on NCBI under  
863 BioProject ID PRJNA1021353 and on the National Genomics Data Center (NGDC)  
864 under BioProject ID PRJCA022010. The genome assembly and their annotations have  
865 been deposited into in Zenodo: <https://doi.org/10.5281/zenodo.8278185>.

## 866 **Code availability**

867 All scripts performed in this study are available on GitHub:

868 [https://github.com/zhouyflab/Polygenetic\\_Basis\\_Seedless\\_Grapes](https://github.com/zhouyflab/Polygenetic_Basis_Seedless_Grapes)

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