Genome Resequencing, Improvement of Variant **Calling, and Population Genomic Analyses Provide** Insights into the Seedlessness in the Genus Vitis

of Korea and [†]Fruit Research Division, National Institute of Horticultural and Herbal Science, Wanju, Jeonbuk 55365, Republic of Korea

ORCID IDs: 0000-0002-5967-4347 (M.-S.K.); 0000-0003-0223-1155 (S.-C.J.)

ABSTRACT The seedlessness of grape derived from stenospermocarpy is one of the most prized traits of table or raisin grapes. It is controlled by a complex genetic system containing one dominant gene and multiple recessive genes. Here, we collected dense variation data from high-depth resequencing data of seeded, seedless, and wild relative grape genomes sequenced to > 37x mean depth. Variant calls were made using a modified variant calling pipeline that was suitable for highly diverse interspecific grape accessions. The modified pipeline enabled us to call several million more variants than the commonly recommended pipeline. The quality was validated by Sanger sequencing data and subsequently supported by the genetic population structure and the phylogenetic tree constructed using the obtained variation data, results of which were generally consistent with known pedigree and taxonomic classifications. Variation data enabled us to confirm a dominant gene and identify recessive loci for seedlessness. Incidentally, we found that grape cultivar Rizamat contains an ancestral chromosomal region of the dominant gene in Sultanina, a predominant seedlessness donor cultivar. Furthermore, we predicted new candidate causal genes including Vitvi01g00455, Vitvi08g01528, and Vitvi18g01237 associated with the recessive seedless-regulating loci, which showed high homology with genes that regulate seed development in Arabidopsis. This study provides fundamental insights relevant to variant calling from genome resequencing data of diverse interspecific hybrid germplasms such as grape and will accelerate future efforts aimed at crop improvement.

Myung-Shin Kim,*^{,1} Youn Young Hur,[†] Ji Hong Kim,* and Soon-Chun Jeong*^{,2} *Bio-Evaluation Center, Korea Research Institute of Bioscience and Biotechnology, Cheongju, Chungbuk 28116, Republic

Grape is one of the most valuable fruit crops and is annually produced from 7.9 million ha globally according to FAOSTAT in 2018 (http:// www.fao.org/faostat/en/#data/QC). While most is processed into

Copyright © 2020 Kim et al.

wine, a significant proportion (\sim 30%) is also destined for fresh consumption (table grape), dried into raisins, or processed into juice. However, \sim 90% of grape produced from \sim 14,000 ha in Korea is consumed as table grape (http://www.krei.re.kr/). Seedlessness is one of the most prized traits of table or raisin grape. Most of the seedless table grape cultivars with known pedigrees are derived from the stenospermocarpic variety Sultanina, also known as Sultanine or Thompson Seedless (Stout 1936; Bouquet and Danglot 1996). The most widely accepted hypothesis proposed for the inheritance of Sultanina-derived stenospermocarpic seedlessness is that the expression of three independently inherited recessive genes is controlled by a dominant regulator gene (Bouquet and Danglot 1996). Molecular markers tightly linked to the dominant locus have been subsequently found and the locus was later named seed development inhibitor (SDI) (Lahogue et al. 1998). Quantitative trait loci (QTL) mapping studies have confirmed the existence of this dominant locus responsible for between 50% and 90% of total phenotypic variance in this trait, depending on the mapping population and trait evaluation

KEYWORDS

Grape resequencing seedless variant calling Vitis vinifera



doi: https://doi.org/10.1534/g3.120.401521

Manuscript received May 17, 2020; accepted for publication July 21, 2020; published Early Online July 21, 2020.

This is an open-access article distributed under the terms of the Creative Commons Attribution 4.0 International License (http://creativecommons.org/ licenses/by/4.0/), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Supplemental material available at figshare: https://doi.org/10.25387/ a3.12581324.

¹Present address: Plant Immunity Research Center, Plant Genomics and Breeding Institute, College of Agriculture and Life Sciences, Seoul National University, Seoul 08826, Republic of Korea.

²Corresponding author: Bio-Evaluation Center, Korea Research Institute of

Bioscience and Biotechnology, Cheongju, Chungbuk 28116, Republic of Korea. E-mail: scjeong@kribb.re.kr.

(Lahogue *et al.* 1998; Doligez *et al.* 2002; Mejía *et al.* 2007). Several minor-effect QTL that could be recessive or modifying genes have also been described in these reports. The *SDI* locus is located on chromosome 18 (Mejía *et al.* 2007). In this region, the MADS-box gene *AGAMOUS-LIKE* 11 (*VviAGL11*) has been proposed as a candidate for *SDI* (Costantini *et al.* 2008; Mejía *et al.* 2011) because the homologous gene in *Arabidopsis* is involved in ovule and carpel development. The direct role of *VviAGL11* in seed morphogenesis has been confirmed by its ectopic expression in the *Arabidopsis SEEDSTICK* mutant (Malabarba *et al.* 2017). An arginine-to-leucine substitution in *VviAGL11* has been postulated to be the major cause of seedlessness in grapevine cultivars (Royo *et al.* 2018). Thus far, however, recessive seedless-regulating genes have not been further elucidated at the molecular level.

Grape was the first fruit species to have its genome completely sequenced (Jaillon et al. 2007). Despite the early availability of a reference genome of Vitis vinifera subsp. vinifera PN40024 (derived from seeded Pinot Noir and close to homozygosity after 6-9 rounds of selfing), grape population genomics has lagged behind other major crop plants (Schreiber et al. 2018) likely because of its high heterozygosity and long generation time. An analysis of approximately one thousand grape accessions using Vitis9kSNP array has revealed that Vitis vinifera L. subsp. vinifera, a domesticated grape species, has maintained high levels of genetic diversity and rapid linkage disequilibrium (LD) decay due to introgression from local wild progenitor Vitis vinifera L. subsp. sylvestris (Gmel.) Hegi during domestication (Myles et al. 2011). Despite a complex network of close pedigree relationships among elite cultivars, first-degree relationships are rare between wine and table grapes and among grapes from geographically distant regions. Recent studies have explored high levels of genomic variation in a few important cultivars (Di Genova et al. 2014; Cardone et al. 2016; Xu et al. 2016; Ma and Yang 2018). A population-level genomics study of grape has investigated the domestication history of grape using genome resequencing data from nine sylvestris and 18 vinifera individuals (Zhou et al. 2017, 2019). However, comprehensive genome resequencing data at a high-depth coverage have only recently been used to investigate the population genomics of grapes to explore grape features other than domestication during the course of this study (Liang et al. 2019; Massonnet et al. 2020).

With an interest in elucidating seedless mechanisms in grape using genome-wide variation data, we have sequenced a diverse group of grape accessions (Hur et al. 2019). In this study, we report analyses of high-depth resequencing data from 33 grape accessions consisting of 14 seeded, 17 seedless, and two wild grape genomes sequenced to a mean depth $> 37 \times$. Such depth is likely sufficient for calling heterozygous genotypes (Ajay et al. 2011). In particular, because many Vitis hybrid cultivars have been generated for various purposes such as disease resistance, environmental adaptation, and flavors and are already cultivated, we obtained resequencing data from several cultivars generated from crosses between V. vinifera and its wild relative species such as Vitis labrusca L. However, the high diversity of our sequenced grape accession required a modification of the recommended popular variant calling pipeline. Data were first used to examine evolutionary relationships between seeded and seedless grapes and determine patterns of population structure and the decay of LD in the seeded and seedless grapes. We then used variation data to understand patterns of nucleotide diversity and LD surrounding the SDI locus and identify the recessive loci underlying seedlessness. The results of this study will be of great value both to grape breeders who are striving to more effectively harness

haplotype variation at seedless-regulating loci to develop superior seedless grape cultivars and to genome researchers in general.

MATERIALS AND METHODS

Sample preparation and sequencing

We collected leaf tissues from a total of 27 individuals consisting of nine seeded, 16 seedless, and two wild grape accessions. Of these, 26 were selected from a grape collection grown in an experimental field of the National Institute of Horticultural and Herbal Science, Wanju, Korea while one accession, Rizamat Gs, was selected from a nursery at Gyeongsan, Korea (Table 1). Rizamat Wj and Rizamat Gs are clones because Rizamat Gs was vegetatively propagated from Rizamat Wj in 1990s. Although we already knew seedless and seeded phenotypes of the selected accessions on the basis of germplasm descriptions available, the phenotypes were verified during the autumns of 2016 through 2018, which were the leaf tissue collection years. Because the main trunk of Rizamat Wj turned out to be dead and a shoot grew out from belowground during the 2018 growing season, we resequenced another Rizamat clone, Rizamat Gs, to confirm the seed phenotype of our resequenced Rizamat cultivar in 2019. Four out of nine seeded grape accessions were interspecific hybrids between V. vinifera and wild grape species. Four of 16 seedless grape accessions were hybrids. Genomic DNA was extracted from each leaf sample using Qiagen DNeasy plant kit. DNA sequencing was performed at Macrogen (Seoul) company in Korea. Paired-end sequencing libraries were constructed with an insert size of 500 bp using TruSeq DNA PCR-Free kit (Illumina, San Diego, CA) according to Illumina library preparation protocols. Libraries were then sequenced using Illumina HiSeq 4000 platform with 2×151 -bp paired reads to a target coverage of 40×. Raw sequencing data were deposited in the Short Read Archive at NCBI (BioProject PRJNA485199). We also used Illumina raw reads with >37 coverage depths for five other seeded and one seedless cultivars (Da Silva et al. 2013; Di Genova et al. 2014; Gambino et al. 2017; Mercenaro et al. 2017; Zhou et al. 2017) that were downloaded from the Short Read Archive at NCBI (Table 1). In results, we ended up with high coverage genome resequencing data of 33 grape accessions.

Sequence alignment and variant calling

Short paired-end reads were quality checked using FastQC (http:// www.bioinformatics.babraham.ac.uk/projects/fastqc/). We then essentially followed procedures described in the Genome Analysis Toolkit (GATK) Best Practices for data pre-processing (DePristo et al. 2011; Van der Auwera et al. 2013) with some modifications. We used BWA (version 0.1.12) with default parameters (Li and Durbin 2009) to map genomic reads from each accession against V. vinifera Pinot Noir PN40024 reference genome (12X.v2) (Canaguier et al. 2017). Alignments were further checked for PCR duplicates using Picard (version 1.134) (http://picard.sourceforge.net/). We performed data pre-processing, including sorting operation and base recalibration, using GATK (version 4.0.1.2). A total of 457,245 known variants of grape genomes used in this study were release-40 data downloaded from https://plants.ensembl.org/index.html (accessed on September 10, 2018). Their coordinates were converted to 12X.v2 coordinates using python script provided by Canaguier et al. (2017) before use. For variant calling, we used GATK (v. 3.8-0) UnifiedGenotyper available in GATK3 because our initial attempt using HaplotypeCaller implemented in GATK4 produced a much smaller number of SNPs from hybrids than from V. vinifera, in contrast to the simple assumption that distantly related accessions would have higher genetic variation than closely related accessions. After further

Table 1 List of sequenced grape accessions and their seed phenotype and mean mapping depth

			de visu		Mapping	GenBank accession	
Accession name	Sample name	Species ^a	seed	Use	depth	code	Reference
Autumn Royal	Autumn_Royal	Vitis vinifera	Seedless	Table	51.99	SAMN09786254	Hur et al. 2019
Autumn Seedless	Autumn_Seedless	V. vinifera	Seedless	Table	59.49	SAMN09786255	Hur <i>et al.</i> 2019
Cheonasoo	Cheonasoo	Vitis sp.	Seedless	Table/wine	51.68	SAMN09786258	Hur et al. 2019
Crimson Seedless	Crimson_SDS	V. vinifera	Seedless	Table	51.44	SAMN09786259	Hur <i>et al.</i> 2019
Dawn Seedless	Dawn_SDS	V. vinifera	Seedless	Table	55.39	SAMN09786260	Hur <i>et al.</i> 2019
Fantasy Seedless	Fantasy_SDS	V. vinifera	Seedless	Table	54.76	SAMN09786261	Hur <i>et al.</i> 2019
Himrod	Himrod	Vitis labrusca x V. vinifera	Seedless	Table/wine	50.13	SAMN09786262	Hur et al. 2019
Hongju	Hongju	V. vinifera	Seedless	Table	48.79	SAMN09786263	Hur <i>et al.</i> 2019
Kishmish Chernyi	Kishm_Chernyi	V. vinifera	Seedless	Table/wine	52.43	SAMN09786265	Hur <i>et al.</i> 2019
Perlon	Perlon	V. vinifera	Seedless	Table	52.35	SAMN09786267	Hur <i>et al.</i> 2019
Princess	Princess	V. vinifera	Seedless	Table	61.84	SAMN09786268	Hur <i>et al.</i> 2019
Ruby Seedless	RSDS	V. vinifera	Seedless	Table	64.32	SAMN09786271	Hur <i>et al.</i> 2019
Ruby Seedless 1	Ruby_SDS_1	V. vinifera	Seedless	Table	48.93	SAMN09786269	Hur <i>et al.</i> 2019
Shiny Star	Shiny_Star	Vitis sp.	Seedless	Table	49.97	SAMN09786272	Hur <i>et al.</i> 2019
Suffolk Red	Suffolk_Red	V. labrusca x V. vinifera	Seedless	Table	47.39	SAMN09786273	Hur et al. 2019
Sultanina		V. vinifera	Seedless	Table/wine	98.04	SAMN02212234	Di Genova <i>et al.</i> 2014
Thompson Seedless	Thompson_SDS	V. vinifera	Seedless	Table/wine	63.81	SAMN09786275	Hur <i>et al.</i> 2019
Bailey Alicante A	Baily_Alic	Vitis sp.	Seeded	Table	54.18	SAMN09786256	Hur <i>et al.</i> 2019
Campbell Early	Campbell_Early	V. labrusca x V. vinifera	Seeded	Table	53.22	SAMN09786257	Hur et al. 2019
Chardonnay 04		V. vinifera	Seeded	Wine	60.28	SAMN07174805	Zhou <i>et al.</i> 2017
Cannonau		V. vinifera	Seeded	Wine	37.57	SAMN07289226	Mercenaro <i>et al.</i> 2017
Italia	Italia	V. vinifera	Seeded	Table/wine	64.67	SAMN09786264	Hur <i>et al.</i> 2019
Muscat of Alexandria	Muscat_Alex	V. vinifera	Seeded	Table/wine	51.71	SAMN09786266	Hur <i>et al.</i> 2019
Nebbiolo_CVT71		V. vinifera	Seeded	Wine	57.48	SAMN07156302	Gambino et al. 2017
Red Globe	RG	V. vinifera	Seeded	Table	56.77	SAMN10995907	This study
Rizamat Gs	RZ	V. vinifera	Seeded	Table	56.23	SAMN10995908	This study
Rizamat Wj	Hur_Seedless	V. vinifera	Seeded		56.38	SAMN09786270	Hur et al. 2019
Tamnara	Tamnara	V. labrusca x V. vinifera	Seeded	Table	51.10	SAMN09786274	Hur et al. 2019
Tannat		V. vinifera	Seeded	Wine	84.13	SAMN02147099	Da Silva et al. 2013
Tano Red	TR 1	Vitis sp.	Seeded	Table	50.91	SAMN09786276	Hur et al. 2019
Zinfandel 03		V. vinifera	Seeded	Wine	41.19	SAMN07174813	Zhou et al. 2017
Vitis amurensis, Cheorwon	V_amurensis	Vitis amurensis	outgroup		50.14	SAMN09786277	Hur <i>et al.</i> 2019
Vitis flexuosa, Chuncheon	V_flexuosa	Vitis flexuosa	outgroup		53.82	SAMN09786278	Hur <i>et al.</i> 2019

^aHybrids among more than three *Vitis* species are presented as *Vitis* sp.

processing by applying IndelRealigner, we called SNPs and indels with UnifiedGenotyper. Raw variant calling data were divided into SNPs and indels with SelectVariants function of GATK (v. 4.0.1.2). Hard-filtering was then performed for these raw SNP calls using VariantFiltration function of GATK (v. 4.0.1.2) according to the following threshold criteria: MappingQualityRankSum of < -12.5, polymorphism confidence scores (QUAL) < 30, genotype call quality divided by depth (QD) < 3.0, Phred-scaled P value of Fisher exact test for strand (FS) >30.0, mapping quality (MQ) < 30.0, total depth of coverage (DP) <200, and genotype-filter-expression depth of coverage (DP) < 15. Bi-allelic variants were then selected using VCFtools (version 0.1.15) (Danecek et al. 2011). To exclude erroneous variants in repetitive regions, variants with high mapping depth (> 4X reads per sample, where X is the mapping depth) in each sample were filtered. Allele balance (AB) was calculated and variants with AB < 30 in heterozygous genotypes were filtered. Variants with AB > 30 in homozygous

genotypes in interspecific hybrids and wild relative species were converted to heterozygous genotypes. SNPs with missing rate > 30% were removed using VCFtools (version 0.1.15) (Danecek *et al.* 2011). Filtering of raw indel calls was performed according to the following threshold criteria: ReadPosRankSum of < -20.0, QUAL < 30, QD < 2.0, and FS > 200, and DP < 200. Bi-allelic variants were then retained.

From this analysis, a total of 17,453,275 filtered SNPs and 3,109,464 filtered indels were defined as candidate variants. To perform population analyses, we further filtered these candidate SNPs using VCFtools (version 0.1.15) (Danecek *et al.* 2011) according to the following criteria: --non-ref-ac 1 --maf 0.05 --max-missing 0.9. Finally, we retained 5,373,452 high-quality SNPs in the data set.

Variant calling with BCFtools

For variant calling, BCFtools (v. 1.9) (https://samtools.github.io/ bcftools/bcftools.html), a variant calling project split from SAMtools package (Li 2011), was used to conduct analysis for output files from the above-described IndelRealigner step of GATK with the following options: bcftools mpileup -Ou -a FORMAT/AD, FORMAT/DP; bcftools call -Ov -mv -f GQ. VariantAnnotator function of GATK (v. 3.8-0) was used to add the following additional annotations: QualByDepth, MappingQualityRankSumTest, and FisherStrand. SNPs were then filtered using VariantFiltration function of GATK (v. 4.0.1.2) with criteria described above.

Validation of variants

We validated candidate SNPs and indels called from genome resequencing data by Sanger sequencing of genomic DNA fragments PCR-amplified from twelve genes in four V. vinifera (Autumn Royal, Honju, Italia, Muscat of Alexandria, Rizamat Gs), two interspecific hybrid (Campbell Early and Cheongsoo), and one wild relative (V. amurensis) accessions. Primer sets were designed to amplify sequences in the genomic region of these genes (Table S1). In particular, three primer sets that would amplify overlapping gene fragments for the assembly of three contigs were designed to amplify sequences in the up-and down-stream and in exon/intron regions of VviAGL11 in order to sequence the whole gene. As a result, the contigs from the same haplotypes were identified by specific polymorphisms in overlapping sequences. PCR amplifications were performed with 20 ng of grape genomic DNA using TAKARA LA Taq (Cat No. RR002A) with annealing temperature of 53° for most of the genes except VviAGL11. Because gene fragments amplified for VviAGL11 are larger than 3 kbp, we performed PCR amplifications using TAKARA LA Taq (Cat No. RR002A) with annealing temperature of 65°. Because these called variants contained an appreciable rate of heterozygous variants, a given PCR product was subcloned into a plasmid for sequencing. At least three different clones for each haplotype were then sequenced. Sequences of both ends of a PCR amplicon were also determined directly from the amplified products to predict copy number of the amplicon based on sequence profiles.

Population structure and relatedness

Population groups were inferred using a Bayesian model-based clustering method, fastSTRUCTURE (version 1.0) (Raj et al. 2014). Fast-Structure was run on default settings on the 33 grape accessions together with the Pinot Noir reference genome. The number of subpopulations (K) ranged from 1 to 12. Python script ChooseK incorporated with the FastStructure package was used to choose the number of subpopulations that could maximize the marginal likelihood. Results were graphically represented using STRUCTUREPLOT v1 (Ramasamy et al. 2014) with the Structure plot ordered by Q-value and accession names included as individual labels. Phylogeny among grape genomes was assessed using the neighbor-joining and bootstrap method implemented in MEGA7 (Kumar et al. 2016). Neighbor-joining trees were generated using *p*-distance measurement, pairwise deletion treatment, and 1000 bootstrap replicates to assess branch support. Principal Component Analysis (PCA) was performed using SMARTPCA with default setting (Patterson et al. 2006). For most of downstream analyses, the Pinot Noir reference genome, one of each of duplicated samples (V. vinifera Rizamat Wj, Ruby Seedless 1, and Sultanina), and two wild relative grapes (V. amurensis and V. flexuosa) were excluded.

Linkage disequilibrium (LD)

A total of 28 grape accessions consisting of 20 *V. vinifera*, and eight hybrids were separated and filtered using VCFtools (version 0.1.15) (Danecek *et al.* 2011) with --keep option and the following

criteria: --non-ref-ac 1--maf 0.1--max-missing 0.9. Un-anchored (chr00), mitochondrial, and plastid sequences were also removed with --not-chr option. LD analysis was performed and plotted using PopLDdecay software (v. 3.4.0) (Zhang *et al.* 2019). Average r^2 of each 100 bp block was plotted using Plot_OnePop.pl script implemented in PopLDdecay.

Predicting variant functional impact with SIFT

To predict functional effects of variants, Sorting Intolerant From Tolerant 4G (SIFT 4G) software (Vaser *et al.* 2016) was used. To create a grape database, uniref90 (https://www.uniprot.org/, download date: Feb 9th, 2019) was used as reference protein set. Annotation of *Vitis vinifera* 12X.v2 was downloaded from URGI (https://urgi.versailles.inra.fr/files/ Vini/Vitis%2012X.2%20annotations/Vitis_vinifera_gene_annotation_ on_V2_20.gff3.zip). Gff3 format was converted to Ensembl GTF format. Grape SIFT 4G database was constructed using SIFT4G_Create_Genomic_ DB implemented in SIFT 4G. Functional effects of variants in coding regions of 33 grapes were predicted using SIFT 4G annotator with default option.

Genome scanning for selective signals

Twenty-eight grapes consisting of 13 seeded and 15 seedless grape accessions after excluding three duplicated samples and two wild relative samples were used for detecting selective sweep regions and logistic association. Monomorphic, minor allele frequency < 5%, and missing rate > 10% markers were filtered using VCFtools (version 0.1.15) (Danecek et al. 2011). Missing variants were imputed using BEAGLE v4.0 (Browning and Browning 2007) with default option. We then performed a genome scan using a composite likelihood approach (XP-CLR) (Chen et al. 2010) updated by Hufford et al. (2012). Although XP-CLR has widely been used to detect selective signals from domestication process in plant researches, it has successfully been used to detect other adaptive introgressions such as environment adaptation (Liu et al. 2015) and crop improvements (Hufford et al. 2012; He et al. 2019). Following this strategy, we screened evidence for selection of seedlessness across the genome by comparing seeded vs. seedless grape genomes. Individual SNPs were assigned at positions along a Ri parental map derived from crosses between V. vinifera cv. Riesling cl.49 and V. vinifera cv. Gewürztraminer cl.643 (Ri ×Gw) downloaded from URGI (https://urgi.versailles.inra.fr/download/vitis/Genetic_maps_Vitis_12X_V2.zip) (Canaguier et al. 2017). Coordinates of 12X.v2 genome assembly were applied using python script provided by Canaguier et al. (2017) to calculate genetic per physical distance between markers in the Ri map. XP-CLR was performed with the following criteria: -w1 0.0005 200 100 -p1 0.7. In other words, XP-CLR scores of 100 bp window were calculated for maximum 200 SNPs per 0.05 cM genetic window and markers with a correlation level > 0.7 were down-weighted. Additionally, we estimated the fixation index (F_{ST}) to measure population differentiation between the seeded and seedless grape populations to investigate the selection signals for seedlessness across the genome. A 100 kb sliding window with 10 kb step approach was applied to quantify F_{ST} with VCFtools (version 0.1.15) (Danecek et al. 2011).

Seeded and seedless traits were encoded as binary traits of 1 (control) and 2 (case), respectively. Case-control logistic mixed model association test was performed using GENESIS (Gogarten *et al.* 2019) with default logistic mixed model association parameters assessed by Shenstone *et al.* (2018). The Manhattan plots of XP-CLR scores and logistic association *p*-values were constructed using qqman (Turner 2018) in R package.

Data availability

Short read data were deposited in the Short Read Archive at NCBI (BioProject PRJNA485199). Large datasets including SNP and indel calls and SIFT data are available from figshare repository (https://figshare.com/projects/Grape_resequencing_seedlessness_project/63569). The Sanger sequencing data from this study have been deposited with the GenBank data library under Accession Nos. MN243829–MN243907. A supplementary material file in the online of this article contains Figures S1 to S10 and Tables S1 to S7. Supplemental material available at figshare: https://doi.org/10.25387/g3.12581324.

RESULTS

Variant calling

We analyzed resequencing data collected from a total of 33 grape accessions consisting of 13 seeded, 18 seedless, and two wild relative species with > 52x genome coverage (raw data) for variant calling (Table 1). Of these 33, we have recently reported 28 genome resequencing data that included eight seeded, 18 seedless grape cultivars, and two wild relatives as outgroups resequenced for this study (Hur *et al.* 2019). For the present study, we resequenced two additional seeded grape cultivars and added data from three previously reported accessions (Da Silva *et al.* 2013; Gambino *et al.* 2017; Mercenaro *et al.* 2017).

V. vinifera divided into subspecies vinifera and its progenitor subspecies sylvestris was the only species of food grape until the end of the 19th century. However, since the outbreak of phylloxera at the end of the 19th century, interspecific hybrids between V. vinifera and other interfertile Vitis species were extensively introduced for disease resistance, different flavors, or adaptation to geographic areas other than the Mediterranean region (This et al. 2006). Because we opted for resequencing of important cultivars that were supposed to be better adapted in the Korean peninsula, we included four seeded and four seedless interspecific hybrids (Table 1). After removing duplicate mapped reads, the mapped mean depth was > 37x for all accessions. More than 92% of the reference genome was covered by more than one read and > 87% were covered by more than five reads for all accessions. Thus, the mapping rate for hybrids and wild relatives is even better than those reported when resequencing data of wild rice Oryza rufipogon were mapped against the rice reference genome sequence from Oryza sativa sub. japonica (Xu et al. 2012). However, when we conducted raw candidate SNP calling using HaplotypeCaller implemented in GATK4 in our initial analysis (Figure 1a), we obtained approximately one million less SNPs from most of the hybrids and wild relatives than those from V. vinifera (Figure S1a). This stands in stark contrast to the simple assumption that distantly related accessions would have higher genetic variation than closely related accessions, when compared to the reference genome sequence. This phenomenon became worse after a VariantFiltration step (Figure S2a). HaplotypeCaller calls SNPs and indels simultaneously via local de-novo assembly of haplotypes in an active region where it remains to be candidate variant loci on the basis of reads mapping through data pre-processing steps (Van der Auwera et al. 2013). When we examined local assembly results using Integrative Genomics Viewer (Thorvaldsdottir et al. 2013), we observed that large portions of mapped reads became inactive after the local assembly, especially for hybrid and wild relative accessions (Figure S3). Local assembly results are used to obtain likelihoods of alleles for each variant in GVCF output files. This difference likely played a significant role at the multi-sample variant calling stage run through GenotypeGVCFs. In results, when we examined distribution of depth of coverage (DP) values from our grape accessions in raw SNP calling data obtained through GenotypeGVCFs, we found that hybrid and wild relative species contained significantly higher levels (several fold higher) of < 15 DP values than *V. vinifera* (Figure 1b). Thus, for variant calling, we opted to use UnifiedGenotyper available in an older version of GATK which is a position-based caller without local re-assembly. Distribution patterns of DP values in raw SNP calling data obtained through UnifiedGenotyper were similar among V. vinifera, hybrids, and wild relative species (Figure 1c). Moreover, the numbers of SNPs from hybrids and wild relative species were higher than those from V. vinifera (Figure S1b and Figure S2b), as expected based on their phylogenetic relationships. We also attempted to use BCFtools for variant calling. However, the number of raw SNPs or the number of filtered SNPs in each grape accession was approximately ten-fold lower than that from UnifiedGenotyper or HaplotypeCaller (Figure S2c). Therefore, we did not use it further.

Differences in variant calling between HaplotypeCaller and UnifiedGenotyper might occur because some sequences from hybrids and wild relative species that are highly diverse relative to the grape reference genome sequence might have been treated as erroneous sequences. To examine this possibility, we compared V. vinifera variation data from only 23 V. vinifera and from all 33 grape accessions called through GenotypeGVCFs, and found no difference at identical sites between the two data sets. We also compared variant data called through GenotypeGVCFs with data called through UnifiedGenotyper from all 33 grape accessions. Interestingly, genotypes of SNPs that had identical coordinates between the two data sets were more than 93% similar in each V. vinifera accession and less than 85% similar in each hybrid and wild relative accession (Figure 1d). These results indicate that UnifiedGenotyper might be more appropriate than HaplotypeCaller (recommended by developer) for variant calling analysis of distant relative species and their hybrids. It should be noted that application of UnifiedGenotyper for analysis of distant grape relative species, which are diploid, is an additional utility of this method because UnifiedGenotyper is recommended only for ploidy or pooled samples.

It was necessary to validate variant calls because the use of UnifiedGenotyper is not recommended by GATK Best Practices (Van der Auwera et al. 2013). To validate variant calling results, we designed primer pairs from genomic regions of eleven randomly selected genes as well as of an SDI-encoding gene, VviAGL11, and performed Sanger sequencing. High-quality sequences determined by multiple clones in ten grape accessions ranged from 5.2 kb to 15.9 kb (Table S2). We also used AGL11 sequences from Chardonnay and Sultanina reported by Malabarba et al. (2017). The sequences contained from 92 to 220 candidate SNP sites called through UnifiedGenotyper. Of the SNPs, 96.2% of UnifiedGenotyper SNP calls could be validated. However, only 77.4% of HaplotypeCaller SNP calls were validated (Table S2). As expected from the high similarity between UnifiedGenotyper and HaplotypeCaller SNP calls for V. vinifera (Figure 1d), 96.2% of UnifiedGenotyper and 96.0% of Haplotype-Caller SNP calls were validated by Sanger sequences for the seven V. vinifera. As also expected from the lower similarity between UnifiedGenotyper and HaplotypeCaller SNP calls for interspecific hybrids and wild relatives other than V. vinifera (Figure 1d), much lower percentages of HaplotypeCaller than those of UnifiedGenotyper SNP calls were validated by Sanger sequences; 83.8% of Unified-Genotyper and 38.6% of HaplotypeCaller SNP calls were validated by Sanger sequences for two hybrids and 92.9% of UnifiedGenotyper and 55.8% of HaplotypeCaller SNP calls were validated for a V. amurensis



Figure 1 Comparison of workflows for variant discovery from grape genome sequencing data. a Workflows of HaplotypeCaller-based variant calling pipeline similar to the GATK Best Practices and UnifiedGenotyper-based variant calling pipeline. b Distribution of the depth of coverage (DP) per site from each of our grape accessions in raw SNP calling data obtained through GenotypeGVCFs using the HaplotypeCaller-based pipeline. *Vitis vinifera*, interspecific grape hybrids, and wild relative species are shown by blue, green, and orange lines. c Distribution of the DP per site from each of our grape accessions in raw SNP calling data obtained through UnifiedGenotyper using the UnifiedGenotyper-based pipeline. *Vitis vinifera*, interspecific grape hybrids, and wild relative species are shown by blue, green, and orange lines. c Distribution of the DP per site from each of our grape accessions in raw SNP calling data obtained through UnifiedGenotyper using the UnifiedGenotyper-based pipeline. *Vitis vinifera*, interspecific grape hybrids, and wild relative species are shown by blue, green, and orange lines. Accessions with relatively higher DP per site are Cannonau (*V. vinifera*) and Suffolk Red (hybrid). d Genotype concordance rate of SNPs called by the GenotypeGVCFs with SNPs called by the UnifiedGenotyper in each of grape accessions. Only the SNPs that shared the grape reference genome coordinates were compared.

accession. This large difference between the two pipelines were mainly due to missing calls of HaplotypeCaller. These findings likely explain the six million more SNPs from UnifiedGenotyper than that from HaplotypeCaller in total SNP calls.

When we examined the variant call format (VCF) file from UnifiedGenotyper, we found that many sites with higher than 30% of allelic balance were called homozygous SNPs. Interestingly, total numbers of SNP sites that were supposed to be erroneously called based on allelic balance were several hundred thousand for each of the hybrid and wild relative species accessions, with the highest number of 1.3 million seen for V. flexuosa (Table S3). However, these SNP sites numbered less than 10,000 for each V. vinifera accession, and these may be considered basal level errors. This phenomenon might have occurred because the UnifiedGenotyper caller purged alleles from more diverse reads of two haplotypes. Thus, we decided to convert these homozygous SNP to heterozygous SNPs. After this conversion, UnifiedGenotyper SNP calls for hybrid and wild relative accessions turned out to be highly accurate: 97.6% (13.8% increase from before conversion) and 92.9% (5.6% increase) of Unified-Genotyper SNP calls were validated by Sanger sequences for hybrids and wild relative accessions, respectively. In sum, we identified in all 33 accessions approximately 17.45 million candidate SNPs that went through the quality control filtering described in the Materials and

Methods section below (Table S4). To obtain high-quality SNPs for population analyses, we excluded SNPs with < 10% minor allele frequency (MAF) and > 10% missing rate, yielding a total of 5,373,452 high-quality SNPs (Table S5). Relative to other accessions, Cannonau and two wild relative accessions consistently showed higher number of missing sites in all filtering methods and steps (Figure S4). In the case of Cannonau, this might occur due to its lowest mapping depth. For the two wild relatives, it might reflect some chromosomal regions that were too diverse for short reads to map properly.

Population structure

A set of 5,373,452 high-quality SNPs was used to examine the genetic population structure and relationships among these 33 grape accessions together with the Pinot Noir reference genome. To analyze the population structure, fastSTRUCTURE program (Raj *et al.* 2014) was used to estimate individual ancestry and admixture proportions assuming that *K* populations existed based on a maximum-likelihood method. The estimated marginal likelihood value plot in this analysis clearly supported the presence of three clusters (Figure S5). At K = 2, grape accessions were divided into two groups consisting of a group of *V. vinifera* and a group of interspecific hybrids and two wild relatives (Figure 2a). At K = 3, *V. vinifera* accessions formed two

groups. Interestingly, V. vinifera accessions were roughly divided into wine and table grapes. These results are consistent with those of previous studies (Myles et al. 2011; Emanuelli et al. 2013) showing that close pedigree relationships are rare between wine and table grapes. Interspecific hybrids represented by Campbell Early and Tamnara formed an independent group together with two wild relative grape species. However, the other hybrids appeared to be admixtures between V. vinifera and wild relative species. These grouping results that are consistent with pedigree and taxonomy indicated overall accuracy of variant calling in the present study. However, these results indicated that the grouping pattern based on SNPs was not related to that of grapes based on seeded and seedless phenotypes, which were well described at the time of cultivar releases as well as verified in this study (Table 1). Principal component analysis (PCA) results using SMARTPCA with default setting (Patterson et al. 2006) were consistent with grouping from the fastSTRUCTURE analysis (Figure 2b and Figure S5).

The genetic population structure and relationships among these 33 resequenced grape accessions were further examined by constructing a neighbor-joining phylogenetic tree (Saitou and Nei 1987; Kumar et al. 2016). Consistent with fastSTRUCTURE results, these 33 grape accessions could be largely divided into three subclades (Figure 3a). Group I contained only V. vinifera accessions except Bailey Alicante A and Suffolk Red, while Group II consisted of all known interspecific hybrids. The two wild relative species formed an outgroup. In Group I, wine grapes clustered separately, as observed in the fastSTRUCTURE results. The grouping of Bailey Alicante A and Suffolk Red with V. vinifera was consistent with results obtained from fastSTRUCTURE analysis showing that ancestry fraction from the V. vinfera group was considerably high in both accessions. Branch lengths among hybrids and wild relative species were longer than those among V. vinifera accessions, indicating that the diversity level of grapes obtained in this study was consistent with collection data in view of taxonomy and pedigree. It is notable that a tree constructed from SNP data without MAF filtration had much longer branch lengths for the two wild relative species than those in the tree with MAF filtration (Figure S6).

To estimate the LD patterns in different V. vinifera and interspecific hybrid groups, we calculated r^2 (Hill and Robertson 1968) between pairs of SNPs using PopLDdecay (Zhang *et al.* 2019). LD decayed to its half-maximum within approximately 11 kb for V. vinifera (Figure S7), which is similar to that for both wild and cultivated V. vinifera previously reported (Zhou *et al.* 2017). However, for interspecific hybrids, LD was high with a half-maximum of over 300 kb, a size that might be expected from a recently established population.

Although seeded and seedless grape accessions were intermixed within these two clearly separated subgroups in the phylogenetic tree constructed using genome-wide SNPs, seedless-regulating chromosomal regions introgressed from ancient seedless cultivar might be confined within the diverse genetic background. To test this hypothesis, we constructed a tree using 1,744 SNPs from a 100 kb region surrounding the well-characterized *SDI* locus coding for *VviAGL11* (Figure 3b). As expected, the tree formed two independent groups (seeded or seedless grape group) with the exception of Rizamat Wj and Rizamat Gs. Branch lengths within the seedless grape group appeared to be much shorter than those within the seeded grape group, supporting the notion of a single origin for the *SDI*-containing chromosomal region. To further examine *VviAGL11* sequences, we sequenced 8.9 kb of genomic DNA containing this gene. Although we had some difficulty due to preferential PCR amplification of parts of

the haplotypes in several grape accessions (Figure S8), we were able to sequence several haplotypes of the full-length VviAGL11 gene including a Sultanina mutant haplotype of Rizamat Gs. The Rizamat Gs VviAGL11 sequence showed two SNPs in non-coding regions and only one SNP (arginine-to-leucine substitution site in VviAGL11) with several indels at the microsatellite repeat regions in non-coding regions. Moreover, the Rizamat Gs sequence grouped with seedless VviAGL11 mutant sequences in our phylogenetic tree (Figure 4). Most grape accessions contained two haplotypes of the VviAGL11 gene on the basis of our phylogenetic tree constructed from upstream and first coding sequences of VviAGL11 whose two haplotypes were PCR-amplified from accessions attempted in this study (Figure S9). However, among the approximately 300 SNPs and indels detected, only eight SNPs were located in coding regions, suggesting that coding regions have been well conserved. Of the eight, only two including the 197 arginine-to-leucine substitution site (Royo et al. 2018) were non-synonymous. The other non-synonymous SNP (210 threonine-to-alanine substitution) was interesting, however it is not likely to be another causal mutation because this site was not detected in our genome-wide logistic association scan described below. These results suggested that the SDI-containing region in our Rizamat clones might be an ancestral sequence where SDI mutation occurred in the Sultanina or its ancestor.

We gathered three pairs of duplicated resequencing data. Ruby Seedless and Ruby Seedless 1 pair was generated due to a plant mislabeling. Rizamat clone pair (Rizamat Wj and Rizamat Gs) was obtained due to a problem with Rizamat Wj and a Thompson Seedless and Sultanina pair was generated after downloading Sultanina resequencing data that were publicly available. Besides the problem for phenotyping of seeds in Rizamat Wj, as Rizamat Wj grouped together with seedless accessions in the tree constructed from the SDI-containing chromosomal region, we opted to obtain resequencing data from another Rizamat clone, Rizamat Gs. The duplicate samples grouped together in our population structure and phylogeny analyses (Figures 2 and 3). Their SNPs were approximately 99% similar to each other, assuring the high quality of our resequencing data. In the following analysis to identify seedless-regulating chromosomal regions, we excluded Ruby Seedless 1, Rizamat Wj, and Sultanina. Additionally, two wild relative species that were distantly grouped with other grape accessions were excluded. Finally, 13 seeded and 15 seedless grape accessions were analyzed. For analysis of this subset of the population, we used slightly lower number of highquality SNPs due to exclusion of SNPs fixed in the subset.

Identification of seedless-regulating chromosomal regions

Population structure and phylogenetic analyses showed that seeded and seedless grape accessions were intermixed within these two clearly separated subgroups. Such results indicate that seedless-regulating chromosomal regions that have undergone artificial selection after introgression might be localized within a diverse genetic background. Selective sweep regions most affected by artificial selection of seedlessness during grape breeding history likely correspond to the one dominant and three recessive genes predicted by genetic analysis (Bouquet and Danglot 1996). However, because two out of the three homozygous recessive genes are sufficient for the expression of the seedless phenotype, selection pressure during breeding for the seedless trait might be weaker for the recessive genes than the dominant gene. To test this hypothesis, we first used a likelihood method, the cross-population composite likelihood ratio XP-CLR (Chen *et al.* 2010) updated by Hufford *et al.* (2012), to scan for extreme allele



Figure 2 Grape population structure. a Population structure of 33 grape accessions together with the reference genome Pinot Noir estimated by fastSTRUCTURE. Each color represents one ancestral population. Each accession is represented by a vertical bar, and the length of colored segment in each vertical bar represents the proportion contributed by ancestral populations. b Principal components of SNP variation in grape accessions using whole-genome SNP data. The plots show the first three principal components. Vitis vinifera, interspecific hybrids, and wild relatives are shown by green, red, and blue dots, respectively.

frequency differentiation over extended linked regions (Figure 5). A total of 30 selective sweeps (Figure 5 and Table S6) were detected in the highest 0.5% of XP-CLR values. Interestingly, one of the major peaks corresponded with the SDI-locus-residing chromosomal region, as a major dominant seedless-regulating QTL reported by numerous studies (Mejía et al. 2007, 2011; Malabarba et al. 2017; Royo et al. 2018). As seedless grape genotypes at the dominant gene are heterozygous or homozygous, a peak from this locus relative to the recessive gene chromosomal region might not be the highest. These results suggested that some selective sweeps detected might correspond to recessive gene regions where three independently inherited recessive genes controlled by the SDI locus reside. When we scanned genome regions with extreme allele frequency differentiation using the estimated F_{ST} values (Figure S10), which are commonly used for measure of population differentiation but may not be optimal for multilocus allele frequency differentiation (Weir et al. 2005; Chen et al. 2010), the results supported our observations from the XP-CLR analysis. We found that the overall chromosomal distribution patterns of both the XP-CLR and F_{ST} values were similar to each other. Most of the major peaks overlapped each other between the two distributions with exceptions that the peaks on chromosome 7 and 16 from the distribution of the XP-CLR values did not appear in the distribution of F_{ST} values.

Although our population size was only 28, consisting of 15 seedless and 13 seeded grape accessions, seedless-regulating SNPs might be more strongly associated with the seedlessness trait than other SNPs. Thus, we attempted to detect SNPs associated with seedlessness using a case-control logistic mixed model association test implemented in GENetic EStimation and Inference in Structured samples (GENESIS) software (Gogarten et al. 2019), with correction of population structure that analyzed a binary phenotype of seeded or seedless phenotype. The highest peak correlated with the SDI locus on chromosome 18, unlike the XP-CLR analysis (Figure S10). Interestingly, an SNP with the highest -log10 P value of 5.051 in the highest peak was the arginine-to-leucine substitution site in VviAGL11 identified as a causal mutation of the SDI locus. Comparison between XP-CLR and logistic association results showed that the majority of peaks were overlapping with each other. However, chromosome 19 contained two high peaks in logistic association scan while it did not contain a selective sweep and, vice versa, the end of chromosome 11 contained no significant peak in logistic association scan, however it did contain a high selective sweep peak. Those non-overlapping peaks between XP-CLR and logistic association scan results might be false positives generated by population structure or kinship. Thus, we focused on examining variants under the overlapping peaks in detail, with an expectation that we might pinpoint candidate causal genes for the postulated recessive genes.

Assessment of variation patterns for causal gene prediction

Our SIFT (Sorting Intolerant From Tolerant) analysis (Vaser *et al.* 2016) predicted that 1,220 SNPs were deleterious in chromosomal regions of 50 kb to either side of the highest XP-CLR points in the 30 candidate selective sweeps detected (Table S6). Of the 1,220 SNPs, 41 SNPs in 34 genes showed $-\log_{10} P$ values higher than 2.5 from our logistic association (Table S7). In their milestone inheritance study, Bouquet and Danglot (1996) have shown that a system of three complementary recessive genes independently inherited is placed



Figure 3 Phylogeny of grape. a Neighbor-joining phylogenetic tree of 33 grape nuclear genomes and Pinot Noir genome constructed using the 5,373,452 high-quality SNPs called from whole genome resequencing data. Accessions in the Neighbor-joining tree are represented by different colors: Vitis vinifera (red), interspecific hybrids (blue), and wild relatives (black). Group of wine grape accessions is indicated to emphasize their unique pedigree in this tree. Seedless accessions are in italic. b The tree constructed using 1,744 SNPs from 100-kb chromosomal region that contains the SDI locus in the central position. Percentages higher than 60 based on 1000 bootstrap replicates are shown above branches. Seeded grape accessions are represented by green letters and seedless accessions represented by purple letters. V. vinifera accessions are in italic.

under the control of a completely dominant regulator gene *SDI*. When the *SDI* gene is heterozygous or homozygous dominant, expression of the seedless phenotype requires a minimum of two genes to be homozygous recessive. According to this model, most of the causal mutation sites at the *SDI* locus for seedless accessions should be homozygous or heterozygous non-reference alleles as Pinot Noir of the grape reference genome sequence accession is seeded, and most of causal mutations at the recessive seedless-regulating genes should be homozygous non-reference alleles.

We first examined variation patterns at a peak from 29.46 Mb to 30.46 Mb on chromosome 18, which includes the SDI locus (Figure 5). Seven SNPs that were predicted to be deleterious using SIFT showed -log₁₀ P values of higher than 2.5 from our logistic association scan (Table S7). Of the seven, genotype distribution of only one SNP (nucleotide position 30,306,458 on chromosome 18) in the seedless and seeded grape population was consistent with that predicted by the inheritance model of seedlessness. This SNP was heterozygous in all 15 seedless accessions tested, reference homozygous in 11 of 13 seeded accessions tested, and not called in 2 of the 13. Interestingly, it was the arginine-to-leucine substitution site identified as a causal mutation in VviAGL11 encoding the SDI locus (Royo et al. 2018). Two of the remaining six SNPs were reference homozygous in one and two of 15 seedless accessions, respectively. One other was heterozygous in two seeded accessions, Italia and Nebbiolo_CVT71. Interestingly, all the remaining six were heterozygous in two Rizamat clones. The results are consistent with a tree constructed from a chromosomal region under this peak that showed clear separation of seeded and seedless grape groups with the exception of Rizamat Gs. Rizamat was developed by a cross of landraces (Katta Kurgan and Parkentskii) in Uzbekistan close to Turkey where Sultanina was collected

(http://www.vivc.de/) (Mirzaev and Djavacynce 2004). Thus, our SNP genotyping and the geographic origin of Rizamat suggest that this grape cultivar contained an ancestral chromosomal region of the *SDI* locus in Sultanina, a predominant seedlessness donor cultivar.

To examine variation patterns in the candidate selective sweeps of recessive seedless-regulating genes, we classified all SNPs predicted to be deleterious into three groups using SIFT program in the candidate regions (Table S7). Group I included SNPs that showed $-\log_{10} P$ values higher than 2.5 from our logistic association and were non-reference homozygous recessive in more than 10 of 15 seedless



Figure 4 Neighbor-joining phylogenetic tree constructed from 12 haplotypes of 8.9 kb-genomic DNA sequences encoding the *VviAGL11* gene. Sultanina mutant and Rizamat Gs haplotype 2 are highlighted by purple letters.



Figure 5 Genome-wide likelihood (XP-CLR) values for selection of seedlessness for seedless grape relatives to seeded grape accessions in 5-kb windows across the genome. The chromosome number is indicated along the x-axis. Chromosomal locations of Vitvi01g00455, Vitvi08g01528,

and Vitvi18g01237 associated with the recessive seedless-regulating genes and AGL11 associated with a dominant gene, which were predicted based on this XP-CLR analysis as well as logistic association and SIFT (Sorting Intolerant From Tolerant) analyses, are indicated by * and \mathbf{v} , respectively.

accessions tested. Group II included SNPs that showed $-\log_{10} P$ values higher than 2.5 from our logistic association and were non-reference homozygous recessive in less than 10 of 15 seedless accessions. Group III, which was excluded from further consideration, included SNPs that showed $-\log_{10} P$ values of lower than 2.5 from our logistic association test. Based on these criteria, 13 SNPs in four candidate selective sweeps were assigned to group I, whereas 21 SNPs in eight candidate selective sweeps were assigned to group II.

Six group I SNPs located within four genes were found at a peak from 4.3 Mb to 5.3 Mb on chromosome 1. Of the four genes, Vitvi01g00455 was annotated as a cytosolic phosphoglucomutase (cPGM), a regulator of seed development in Arabidopsis (Egli et al. 2010) (Figure 6a and Table S7). Loss of cPGM in Arabidopsis compromises male and female gametophyte development. Thus, Vitvi01g00455 appears to be a candidate gene at this peak for a recessive seedless-regulating gene. A phylogenetic tree based on 1,669 SNPs from a 100-kb region surrounding Vitvi01g00455 clearly separated seedless grape accessions with the non-reference homozygous recessive SNP and seedless and seeded grape accessions with the other genotypes (Figure 6d). Only one group I SNP at the Vitvi08g01528 gene model was found at a peak from 17.47 Mb to 28.47 Mb on chromosome 8 (Figure 6b and Table S7). Vitvi08g01528 was annotated as a basic helix-loop-helix transcription factor, a homolog of Arabidopsis RETARDED GROWTH OF EMBRYO1 (RGE1) (Kondou et al. 2008). Arabidopsis RGE1 functions as a positive regulator in the endosperm at the heart stage of embryo development and exhibits pleiotropic phenotypes including small shriveled seeds and retardation of embryo growth. This annotation result suggests that Vitvi08g01528 is the best candidate gene for a recessive seedlessregulating gene among the group I SNPs. A phylogenetic tree based on 939 SNPs from a 100-kb region surrounding Vitvi08g01528 clearly separated seedless grape accessions with the non-reference homozygous recessive SNP and seedless and seeded grape accessions with the other genotypes (Figure 6e). Three group I SNPs at the Vitvi08g02370 gene model were found at a peak from 20.20 Mb to 21.20 Mb on chromosome 8. This gene was annotated as a retrotransposon-related gene. Because this short gene with a coding region of 204 bp also has three deleterious SNPs, it is likely a pseudogene. On chromosome 18, the candidate peak was predicted to be from 12.83 Mb to 13.83 Mb. We assigned three SNPs to group I. They were mapped to three genes. Of the three genes, Vitvi18g01230, which is a short gene with a 183-bp coding region, was annotated as a retrotransposon-related gene. Vitvi18g01245 was annotated as a retrotransposon-related probable LRR receptor-like serine/threonine-protein kinase with only one intron. Vitvi18g01237 was annotated as a pentatricopeptide repeat protein, a homolog of Arabidopsis MEF12. Arabidopsis MEF12 is involved in RNA editing in mitochondria (Härtel et al. 2013) and has not been studied for seed development. Considering the importance of RNA editing in plant development, Vitvi18g01237 is a candidate

for a recessive seedless-regulating gene. A phylogenetic tree based on 871 SNPs from 100-kb region surrounding *Vitvi18g01237* clearly separated seedless grape accessions with the non-reference homozygous recessive SNP and seedless and seeded grape accessions with the other genotypes (Figure 6f). Taken together, the grouping pattern in three trees constructed from candidate recessive gene peaks is consistent with the inheritance model of seedlessness and indicates selection of this candidate selective sweep for seedlessness. This indicated that all three genes, which are known to be involved in seed development, are candidate causal genes for the recessive seedless-regulating gene.

Most of the group II SNPs appeared to be homozygous nonreference genotypes in less than five of 15 seedless accessions tested, thereby excluding the possibility of the SNPs-carrying genes for recessive seedless-regulating candidate genes. Five group II SNPs at five genes showed homozygous non-reference genotypes in seven or eight of 15 seedless accessions tested. Of the five genes, only Vitvi08g01518 has been implicated in the process of seed development: its Arabidopsis homolog HD2B (At5g22650) functions as a genetic factor associated with seed dormancy. However, it is unlikely to be one of the recessive seedless-regulating candidate genes because the candidate Vitvi08g01528 with a group I SNP resides at the same candidate selective sweep. The 41 SNP sites selected tended to show homozygous non-reference genotype in none of 13 seeded accessions tested. Several SNPs were non-reference homozygous in only one or two of 13 seeded accessions tested. Interestingly, Rizamat Gs showed homozygous non-reference genotype in all six selected SNPs on chromosome 1. In the phylogenetic tree constructed from the chromosomal region (Figure 6d), Rizamat clones grouped together with seedless grape accessions. The results support our notion that Rizamat has carried an ancestral form of the SDI-residing chromosomal region in Sultanina.

DISCUSSION

Development of variant calling methods from genome resequencing data have typically revolved around humans, livestock animals, and major crop plants, most of which contain diploid genomes. Numerous variant callers including UnifiedGenotyper and SAMtools have been proposed for variant calling of polyploidy species that are prevalent in the plant kingdom. Although none of these software packages is definitively recommended over others, many studies have successfully used the called variant data to address important biological questions (Clevenger *et al.* 2015). However, efforts to call variants in hybrids between relative species and in wild relative species that widely exist in woody species such as grape have been poorly done. Most grape resequencing data have been obtained from *V. vinifera*, which is the species used to obtain the reference genome sequence. Genome resequencing data of 472 *Vitis* accessions published while conducting this study included 108 wild relative grape



Figure 6 Genetic features for the candidate causal genes for recessive seedless-regulating candidate selective sweeps. Coding sequence structures of *Vitvi01g00455* (a), *Vitvi08g01528* (b), and *Vitvi01g01237* (c) and genotype distributions of candidate causal SNPs in these genes for 15 seedless and 13 seeded grape accessions tested. Reference homozygous genotype is indicated by 0/0, heterozygous 0/1, non-reference homozygous 1/1, and missing ./.. Amino acid positions are indicated by numbers. Neighbor-joining phylogenetic trees of 17 seedless and 14 seeded accessions constructed using 1,744 SNPs, 939 SNPs, and 871 SNPs from 100-kb chromosomal regions that contains the *Vitvi01g00455* (d), *Vitvi08g01528* (e), and *Vitvi01g01237* (f) genes in their central position, respectively. Percentages higher than 60 based on 1000 bootstrap replicates are shown above branches. Seeded grape accessions are in green, seedless accessions in purple, and two wild relatives in black.

species and 109 interspecific hybrid accessions. However, variants called using HaplotypeCaller were not validated by an experimental approach. In this study, we show that UnifiedGenotyper is better than HaplotypeCaller for SNP calling in hybrids and wild relative species. We also added an additional step to convert erroneously called homozygous SNPs to heterozygous SNPs based on allelic balance values. Therefore, our modified variant calling pipeline should provide insight for improvement of current variant callers to facilitate molecular genetic studies including marker-trait association studies for interspecific hybrids and wild relative species.

Several lines of evidence suggest that our predicted recessive genetic loci are likely real. First of all, both XP-CLR and logistic association peaks with a shared high peak at the well-characterized *SDI* locus showed that the majority of peaks detected were overlapping. Trees constructed from peak-residing chromosomal regions tended to separate seeded and seedless grape accessions. Several of 30 peaks were located at the same physical locations as previously reported for minor QTL identified using seedless phenotypes in full-sibling F1 populations (Mejía et al. 2007; Costantini et al. 2008). Genotype distributions of deleterious substitutions in genes that reside at three overlapping peaks were supported by those proposed by Bouquet and Danglot (1996) in their milestone inheritance study of seedlessness. As results, we were able to suggest three promising candidate causal genes, namely Vitvi01g00455, Vitvi08g01528, and Vitvi18g01237, as associated with the recessive seedless-regulating genes. It is difficult to pinpoint a causal gene underlying a weak recessive gene even with a large segregating population. In this study, based on analysis of highdensity genome-wide SNP data, we have pinpointed several good candidate seedless-regulating genes that can be tested using techniques such as mutagenesis, transformation, and gene editing in the near future. This was made possible due to millions of high-quality genetic variants detected using our modified variant calling pipeline. In this study, we have provided a large genome-wide variation

dataset for seedless and seeded grape accessions with diverse genetic backgrounds. Because our initial variant calling efforts suggested that the current widely used variant calling pipeline had problems with interspecific hybrids and wild relative species, we modified the pipeline. Variation data from the modified pipeline were validated by Sanger sequencing. Our population structure and phylogenetic analysis using the resultant high-quality SNPs strongly supported known pedigree information as well as taxonomic grouping of these sequenced grape accessions, indicating that our modified pipeline was sound. The resulting millions of high-quality variations also provided an opportunity both to validate a dominant seedless-regulating gene and to predict recessive seedless-regulating genes. Investigation of variation patterns at significant peaks allowed us to predict candidate causal genes that could regulate the seedless trait. Taken together, data generated in this study represent such a diverse grape genome background. They can now be used as dense markers of genome variation for marker-assisted mapping of important grape traits as well as for pinpointing agronomically important genes in grapes.

ACKNOWLEDGMENTS

We thank Dong Jun Im from National Institute of Horticultural and Herbal Science for assistance with collection of grape samples. We would also like to thank two anonymous reviewers for their thoughtful comments and suggestions. This research was supported by a grant (PJ01242104) of the Rural Development Administration, Republic of Korea and by the Korea Research Institute of Bioscience and Biotechnology Research Initiative Program.

LITERATURE CITED

- Ajay, S. S., S. C. Parker, H. O. Abaan, K. V. Fajardo, and E. H. Margulies, 2011 Accurate and comprehensive sequencing of personal genomes. Genome Res. 21: 1498–1505. https://doi.org/10.1101/ gr.123638.111
- Bouquet, A., and Y. Danglot, 1996 Inheritance of seedlessness in grape (*Vitis vinifera* L.). Vitis 35: 35-42.
- Browning, S. R., and B. L. Browning, 2007 Rapid and accurate haplotype phasing and missing-data inference for whole-genome association studies by use of localized haplotype clustering. Am. J. Hum. Genet. 81: 1084–1097. https://doi.org/10.1086/521987
- Canaguier, A., J. Grimplet, G. Di Gaspero, S. Scalabrin, E. Duchene *et al.*, 2017 A new version of the grapevine reference genome assembly (12X.v2) and of its annotation (VCost.v3). Genom. Data 14: 56–62. https:// doi.org/10.1016/j.gdata.2017.09.002

- Cardone, M. F., P. D'Addabbo, C. Alkan, C. Bergamini, C. R. Catacchio et al., 2016 Inter-varietal structural variation in grapevine genomes. Plant J. 88: 648–661. https://doi.org/10.1111/tpj.13274
- Chen, H., N. Patterson, and D. Reich, 2010 Population differentiation as a test for selective sweeps. Genome Res. 20: 393–402. https://doi.org/ 10.1101/gr.100545.109
- Clevenger, J., C. Chavarro, S. A. Pearl, P. Ozias-Akins, and S. A. Jackson, 2015 Single nucleotide polymorphism identification in polyploids: A review, example, and recommendations. Mol. Plant 8: 831–846. https:// doi.org/10.1016/j.molp.2015.02.002
- Costantini, L., J. Battilana, F. Lamaj, G. Fanizza, and M. S. Grando, 2008 Berry and phenology-related traits in grapevine (*Vitis vinifera* L.): from quantitative trait loci to underlying genes. BMC Plant Biol. 8: 38. https://doi.org/10.1186/1471-2229-8-38
- Danecek, P., A. Auton, G. Abecasis, C. A. Albers, E. Banks et al., 2011 The variant call format and VCFtools. Bioinformatics 27: 2156–2158. https:// doi.org/10.1093/bioinformatics/btr330
- Da Silva, C., G. Zamperin, A. Ferrarini, A. Minio, A. Dal Molin *et al.*, 2013 The high polyphenol content of grapevine cultivar tannat berries is conferred primarily by genes that are not shared with the reference genome. Plant Cell 25: 4777–4788. https://doi.org/10.1105/tpc.113.118810
- DePristo, M. A., E. Banks, R. Poplin, K. V. Garimella, J. R. Maguire *et al.*, 2011 A framework for variation discovery and genotyping using nextgeneration DNA sequencing data. Nat. Genet. 43: 491–498. https:// doi.org/10.1038/ng.806
- Di Genova, A., A. M. Almeida, C. Munoz-Espinoza, P. Vizoso, D. Travisany et al., 2014 Whole genome comparison between table and wine grapes reveals a comprehensive catalog of structural variants. BMC Plant Biol. 14: 7. https://doi.org/10.1186/1471-2229-14-7
- Doligez, A., A. Bouquet, Y. Danglot, F. Lahogue, S. Riaz et al., 2002 Genetic mapping of grapevine (Vitis vinifera L.) applied to the detection of QTLs for seedlessness and berry weight. Theor. Appl. Genet. 105: 780–795. https://doi.org/10.1007/s00122-002-0951-z
- Egli, B., K. Kolling, C. Kohler, S. C. Zeeman, and S. Streb, 2010 Loss of cytosolic phosphoglucomutase compromises gametophyte development in *Arabidopsis*. Plant Physiol. 154: 1659–1671. https://doi.org/10.1104/ pp.110.165027
- Emanuelli, F., S. Lorenzi, L. Grzeskowiak, V. Catalano, M. Stefanini *et al.*, 2013 Genetic diversity and population structure assessed by SSR and SNP markers in a large germplasm collection of grape. BMC Plant Biol. 13: 39. https://doi.org/10.1186/1471-2229-13-39
- Gambino, G., A. Dal Molin, P. Boccacci, A. Minio, W. Chitarra *et al.*,
 2017 Whole-genome sequencing and SNV genotyping of "Nebbiolo" (*Vitis vinifera* L.) clones. Sci. Rep. 7: 17294. https://doi.org/10.1038/ s41598-017-17405-y
- Gogarten, S. M., T. Sofer, H. Chen, C. Yu, J. A. Brody *et al.*, 2019 Genetic association testing using the GENESIS R/Bioconductor package. Bioinformatics 35: 5346–5348.
- Härtel, B., A. Zehrmann, D. Verbitskiy, and M. Takenaka, 2013 The longest mitochondrial RNA editing PPR protein MEF12 in *Arabidopsis thaliana* requires the full-length E domain. RNA Biol. 10: 1543–1548. https:// doi.org/10.4161/rna.25484
- He, F., R. Pasam, F. Shi, S. Kant, G. Keeble-Gagnere *et al.*, 2019 Exome sequencing highlights the role of wild-relative introgression in shaping the adaptive landscape of the wheat genome. Nat. Genet. 51: 896–904. https:// doi.org/10.1038/s41588-019-0382-2
- Hill, W. G., and A. Robertson, 1968 Linkage disequilibrium in finite populations. Theor. Appl. Genet. 38: 226–231. https://doi.org/10.1007/ BF01245622
- Hufford, M. B., X. Xu, J. Van Heerwaarden, T. Pyhäjärvi, J. M. Chia et al., 2012 Comparative population genomics of maize domestication and improvement. Nat. Genet. 44: 808–811. https://doi.org/10.1038/ng.2309
- Hur, Y. Y., M. S. Kim, and S. C. Jeong, 2019 Toward identification of haplotypes that control seedlessness of grape by genome resequencing. Acta Hortic. (1248): 171–178. https://doi.org/10.17660/ActaHortic.2019.1248.25
- Jaillon, O., J. M. Aury, B. Noel, A. Policriti, C. Clepet et al., 2007 The grapevine genome sequence suggests ancestral hexaploidization in major

angiosperm phyla. Nature 449: 463–467. https://doi.org/10.1038/ nature06148

- Kondou, Y., M. Nakazawa, M. Kawashima, T. Ichikawa, T. Yoshizumi et al., 2008 RETARDED GROWTH OF EMBRYO1, a new basic helix-loophelix protein, expresses in endosperm to control embryo growth. Plant Physiol. 147: 1924–1935. https://doi.org/10.1104/pp.108.118364
- Kumar, S., G. Stecher, and K. Tamura, 2016 MEGA7: Molecular Evolutionary Genetics Analysis version 7.0 for bigger datasets. Mol. Biol. Evol. 33: 1870–1874. https://doi.org/10.1093/molbev/msw054
- Lahogue, F., P. This, and A. Bouquet, 1998 Identification of a codominant scar marker linked to the seedlessness character in grapevine. Theor. Appl. Genet. 97: 950–959. https://doi.org/10.1007/s001220050976
- Li, H., 2011 A statistical framework for SNP calling, mutation discovery, association mapping and population genetical parameter estimation from sequencing data. Bioinformatics 27: 2987–2993. https://doi.org/10.1093/ bioinformatics/btr509
- Li, H., and R. Durbin, 2009 Fast and accurate short read alignment with Burrows-Wheeler transform. Bioinformatics 25: 1754–1760. https:// doi.org/10.1093/bioinformatics/btp324
- Liang, Z., S. Duan, J. Sheng, S. Zhu, X. Ni *et al.*, 2019 Whole-genome resequencing of 472 Vitis accessions for grapevine diversity and demographic history analyses. Nat. Commun. 10: 1190. https://doi.org/10.1038/ s41467-019-09135-8
- Liu, K. J., E. Steinberg, A. Yozzo, Y. Song, M. H. Kohn et al., 2015 Interspecific introgressive origin of genomic diversity in the house mouse. Proc. Natl. Acad. Sci. USA 112: 196–201. https://doi.org/10.1073/pnas.1406298111
- Ma, Q., and J. Yang, 2018 Transcriptome profiling and identification of functional genes involved in H₂S response in grapevine tissue cultured plantlets. Genes Genomics 40: 1287–1300. https://doi.org/10.1007/ s13258-018-0723-z
- Malabarba, J., V. Buffon, J. E. A. Mariath, M. L. Gaeta, M. C. Dornelas et al., 2017 The MADS-box gene Agamous-like 11 is essential for seed morphogenesis in grapevine. J. Exp. Bot. 68: 1493–1506. https://doi.org/ 10.1093/jxb/erx025
- Massonnet, M., N. Cochetel, A. Minio, A. M. Vondras, J. Lin *et al.*, 2020 The genetic basis of sex determination in grapes. Nat. Commun. 11: 2902. https://doi.org/10.1038/s41467-020-16700-z
- Mejía, N., M. Gebauer, L. Muñoz, N. Hewstone, C. Muñoz *et al.*,
 2007 Identification of QTLs for seedlessness, berry size, and ripening data in a seedless x seedless table grape progeny. Am. J. Enol. Vitic. 58: 499–507.
- Mejía, N., B. Soto, M. Guerrero, X. Casanueva, C. Houel *et al.*,
 2011 Molecular, genetic and transcriptional evidence for a role of *VvAGL11* in stenospermocarpic seedlessness in grapevine. BMC Plant Biol. 11: 57. https://doi.org/10.1186/1471-2229-11-57
- Mercenaro, L., G. Nieddu, A. Porceddu, M. Pezzotti, and S. Camiolo, 2017 Sequence polymorphisms and structural variations among four grapevine (*Vitis vinifera* L.) cultivars representing Sardinian agriculture. Front. Plant Sci. 8: 1279. https://doi.org/10.3389/fpls.2017.01279
- Mirzaev, M. M., and U. M. Djavacynce, 2004 The Schroeder Institute in Uzbekistan: Breeding and germplasm collections. HortScience 39: 917–921. https://doi.org/10.21273/HORTSCI.39.5.917
- Myles, S., A. R. Boyko, C. L. Owens, P. J. Brown, F. Grassi *et al.*, 2011 Genetic structure and domestication history of the grape. Proc. Natl. Acad. Sci. USA 108: 3530–3535. https://doi.org/10.1073/pnas.1009363108
- Patterson, N., A. L. Price, and D. Reich, 2006 Population structure and eigenanalysis. PLoS Genet. 2: e190. https://doi.org/10.1371/ journal.pgen.0020190
- Raj, A., M. Stephens, and J. K. Pritchard, 2014 fastSTRUCTURE: variational inference of population structure in large SNP data sets. Genetics 197: 573–589. https://doi.org/10.1534/genetics.114.164350

- Ramasamy, R. K., S. Ramasamy, B. B. Bindroo, and V. G. Naik, 2014 STRUCTURE PLOT: a program for drawing elegant STRUCTURE bar plots in user friendly interface. Springerplus 3: 431. https://doi.org/ 10.1186/2193-1801-3-431
- Royo, C., R. Torres-Perez, N. Mauri, N. Diestro, J. A. Cabezas *et al.*, 2018 The major origin of seedless grapes Is associated with a missense mutation in the MADS-box gene *VviAGL11*. Plant Physiol. 177: 1234–1253. https:// doi.org/10.1104/pp.18.00259
- Saitou, N., and M. Nei, 1987 The neighbor-joining method: a new method for reconstructing phylogenetic trees. Mol. Biol. Evol. 4: 406–425.
- Schreiber, M., N. Stein, and M. Mascher, 2018 Genomic approaches for studying crop evolution. Genome Biol. 19: 140. https://doi.org/10.1186/ s13059-018-1528-8
- Shenstone, E., J. Cooper, B. Rice, M. Bohn, T. M. Jamann *et al.*, 2018 An assessment of the performance of the logistic mixed model for analyzing binary traits in maize and sorghum diversity panels. PLoS One 13: e0207752. https://doi.org/10.1371/journal.pone.0207752
- Stout, A. B., 1936 Seedlessness in grapes. New York State Agric. Exp. Stn. Tech. Bull. no. 238. New York State Agricultural Experiment Station, Geneva
- This, P., T. Lacombe, and M. R. Thomas, 2006 Historical origins and genetic diversity of wine grapes. Trends Genet. 22: 511–519. https://doi.org/ 10.1016/j.tig.2006.07.008
- Thorvaldsdottir, H., J. T. Robinson, and J. P. Mesirov, 2013 Integrative Genomics Viewer (IGV): high-performance genomics data visualization and exploration. Brief. Bioinform. 14: 178–192. https://doi.org/10.1093/ bib/bbs017
- Turner, S. D., 2018 qqman: an R package for visualizing GWAS results using Q-Q and manhattan plots. J. Open Source Softw. 3: 731. https://doi.org/ 10.21105/joss.00731
- Van der Auwera, G. A., M. O. Carneiro, C. Hartl, R. Poplin, G. Del Angel et al., 2013 From FastQ data to high confidence variant calls: the Genome Analysis Toolkit best practices pipeline. Curr. Protoc. Bioinforma. 43: 11.10.1–11.10.33. https://doi.org/10.1002/0471250953.bi1110s43
- Vaser, R., S. Adusumalli, S. N. Leng, M. Sikic, and P. C. Ng, 2016 SIFT missense predictions for genomes. Nat. Protoc. 11: 1–9. https://doi.org/ 10.1038/nprot.2015.123
- Weir, B. S., L. R. Cardon, A. D. Anderson, D. M. Nielsen, and W. G. Hill, 2005 Measures of human population structure show heterogeneity among genomic regions. Genome Res. 15: 1468–1476. https://doi.org/ 10.1101/gr.4398405
- Xu, Y., Z. Gao, J. Tao, W. Jiang, S. Zhang *et al.*, 2016 Genome-wide detection of SNP and SV variations to reveal early ripening-related genes in grape. PLoS One 11: e0147749. https://doi.org/10.1371/journal.pone.0147749
- Xu, X., X. Liu, S. Ge, J. D. Jensen, F. Hu *et al.*, 2012 Resequencing 50 accessions of cultivated and wild rice yields markers for identifying agronomically important genes. Nat. Biotechnol. 30: 105–111. https:// doi.org/10.1038/nbt.2050
- Zhang, C., S. S. Dong, J. Y. Xu, W. M. He, and T. L. Yang, 2019 PopLDdecay: a fast and effective tool for linkage disequilibrium decay analysis based on variant call format files. Bioinformatics 35: 1786–1788. https://doi.org/ 10.1093/bioinformatics/bty875
- Zhou, Y., M. Massonnet, J. S. Sanjak, D. Cantu, and B. S. Gaut, 2017 Evolutionary genomics of grape (*Vitis vinifera* ssp. vinifera) domestication. Proc. Natl. Acad. Sci. USA 114: 11715–11720. https://doi.org/ 10.1073/pnas.1709257114
- Zhou, Y., A. Minio, M. Massonnet, E. Solares, Y. Lv et al., 2019 The population genetics of structural variants in grapevine domestication. Nat. Plants 5: 965–979. https://doi.org/10.1038/s41477-019-0507-8

Communicating editor: M. Hufford