RESEARCH ARTICLE

Open Access

Super pan-genome reveals extensive genomic variations associated with phenotypic divergence in *Actinidia*

Xiaofen Yu^{1,2}, Minghao Qu^{1,3}, Pan Wu¹, Miao Zhou^{1,3}, Enhui Lai^{1,3}, Huan Liu^{1,4}, Sumin Guo¹, Shan Li¹, Xiaohong Yao^{1*} and Lei Gao^{1,2*}¹⁰

Abstract

Kiwifruit is an economically and nutritionally important horticultural fruit crop worldwide. The genomic data of several kiwifruit species have been released, providing an unprecedented opportunity for pan-genome analysis to comprehensively investigate the inter- and intra-species genetic diversity and facilitate utilization for kiwifruit breeding. Here, we generated a kiwifruit super pan-genome using 15 high-quality assemblies of eight *Actinidia* species. For genebased pan-genome, a total of 61,465 gene families were identified, and the softcore and dispensable genes were enriched in biological processes like response to endogenous stimulus, response to hormone and cell wall organization or biogenesis. Then, structural variations (SVs) against *A. chinensis* 'Donghong' were identified and then used to construct a graph-based genome. Further population-scale SVs based on resequencing data from 112 individuals of 20 species revealed extensive SVs which probably contributed to the phenotypic diversity among the *Actinidia* species. SV hotspot regions were found contributed to environmental adaptation. Furthermore, we systematically identified resistance gene analogs (RGAs) in the 15 assemblies and generated a pan-RGA dataset to reveal the diversity of genes potentially involved in disease resistance in *Actinidia*. The pan-genomic data obtained here is useful for evolutionary and functional genomic studies in *Actinidia*, and facilitates breeding design.

Keywords Kiwifruit, Pan-genome, Actinidia, Structural variation, Disease resistant gene

Core

We present a super pan-genome of eight *Actinidia* (kiwifruit) species. More than one million structural variations were identified among different *Actinidia* species,

*Correspondence: Xiaohong Yao yaox@wbgcas.cn Lei Gao leigao@wbgcas.cn

¹ State Key Laboratory of Plant Diversity and Specialty Crops, Wuhan Botanical Garden, Chinese Academy of Sciences, Wuhan 430074, Hubei, China

² Hubei Hongshan Laboratory, Wuhan 430070, Hubei, China

³ University of Chinese Academy of Sciences, Beijing 100049, China

⁴ Bioinformatics Center, College of Plant Science and Technology, Beijing University of Agriculture, Beijing, China and SVs potentially involved in the phenotypic divergence and environmental adaptation in *Actinidia* were revealed. In addition, a pan-RGA (resistance gene analog) dataset was created to explore the disease resistance gene reservoir in *Actinidia*. This study offers valuable insights into the genomic dynamics in *Actinidia* and valuable resources for kiwifruit improvement.

Gene & Accession Numbers

Sequence reads and the genome assembly with annotation were downloaded from China National Genomics Data Center, Figshare database, GitHub and National Center for Biotechnology Information (Table S1, S2, S8, S13).



© The Author(s) 2025, corrected publication 2025. **Open Access** This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons licence, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons licence and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit http://creativecommons.org/licenses/by/4.0/. The Creative Commons Public Domain Dedication waiver (http://creativecommons.org/publicdomain/zero/1.0/) applies to the data made available in this article, unless otherwise stated in a credit line to the data.

Introduction

Kiwifruit, also known as Chinese gooseberry, is an economically and nutritionally important fruit crop enriched with vital nutrients, especially vitamin C. The global kiwifruit production amounted to approximately 4.54 million metric tons (www.statista.com), and was valued at 1.71 billion dollars in 2022. Kiwifruit is a functional dioecious woody vine belonging to the genus Actinidia of Actinidiaceae. According to recent revision, there are 54 species and 21 varieties identified in Actinidia, which widely distribute throughout eastern Asia ranging from the tropics (latitude 0°) to cold temperate regions (50°N) (Huang 2016). Extensive interspecific hybridizations have driven reticulate speciation and diversification in Actinidia (Liu et al. 2017; Yu et al. 2023). Besides, Actinidia species exhibit complex patterns of ploidy variations with a high basic chromosome number (x=29). For example, there are diploids, tetraploids and hexaploids found in A. chinensis complex, which comprise the two most commercialized kiwifruit varieties, A. chinensis var. chinensis and A. chinensis var. deliciosa. Subdivision of Actinidia has long been contentious, and classification based on morphological features was not sustainable and widely acceptable. Li et al. (2000) proposed to divide Actinidia into two subgenera, Leiocarpae and Maculatae. Leiocarpae, with smooth-skinned, hairless fruit, was consistently formed a monophyletic group. Nevertheless, Maculatae was found polyphyletic which contains individual species varying in frequency of lenticels on the fruit skin and type of leaf hair (Huang 2016). Further subdivision of Maculatae was always paradoxical based on morphological and limited molecular evidence. Since genomics-based approaches have proven their power on disentangling elusive phylogenic relationships (Guo et al. 2023; Li et al. 2024a; Wu et al. 2023), these methods should be adopted to elucidate the complicated interspecific relationship within Actinidia.

Abundant phenotypic variations were identified among Actinidia species. As mentioned above, species in Leiocarpae all have glabrous fruit and almost no pubescence on petiole, while Maculatae species have spotted fruit and indumentum on small flowering shoot (Huang et al. 1999). Additionally, species in Maculatae also have difference in pericarp hair, e.g. A. eriantha with long, straight, and bushy trichomes and A. latifolia with short, distorted, and spare trichomes. Previous study showed that splicing of Nck-associated protein 1 (NAP1) in A. *latifolia* might be responsible for the shorter trichomes in A. latifolia (Miao et al. 2023). However, the genetic basis for pericarp hair variation between Leiocarpae and Maculatae has not been understood so far. Interestingly, some species in Leiocarpae, like A. arguta, has a very short softening stage during fruit ripening comparing to species in Maculatae, e.g. A. chinensis and A. eriantha (Lu et al. 2024). Fruit softening involves change of interactions between pectin, xyloglucan and cellulose, and results in breakdown of the cell wall structure. It was reported that galactose loss as well as pectin solubilization started at an earlier firmness stage in A. arguta fruits compared to A. chinensis var. deliciosa fruits (Sutherland et al. 2017). Besides, A. eriantha and A. latifolia contain higher vitamin C content in fruits than other species (Huang et al. 2000), and the expression level of GGP3 gene was associated with ascorbic acid (AsA) content in A. eriantha (Liu et al. 2021, 2022). A recent study showed that duplication of ERF098 transcription factor in A. latifolia and A. eriantha might account for AsA biosynthesis and accumulation in their fruits (Han et al. 2023). These studies expounded the genetic basis for phenotypic variants in one or two Actinidia speices, while few studies paid attentions to structural variations (SVs) or presence/ absence variants (PAVs), which were found responsible for many phenotypic variations in other species (Qin et al. 2021; Lyu et al. 2023).

It is worth mentioning that, comparing to the cultivated kiwifruits, some wild relatives are more resistant to disease (Wang et al. 2019; Wang et al. 2020a). Kiwifruit industry is threatened by many severe diseases, including bacterial canker, ripe rot, black spot, gray mold and so on (Erper et al. 2013; Pereira et al. 2021; Li et al. 2022; Yang et al. 2022; Zhao et al. 2023). Among them, kiwifruit bacterial canker, caused by Pseudomonas syringae pv. actinidiae (Psa), has posed a serious threat to global kiwifruit industry. Psa damages both A. chinensis var. chinensis and A. chinensis var. deliciosa, resulting in severe economic losses (Pereira et al. 2021). Evaluation on the resistance to Psa in wild Actinidia germplasm showed that some accessions from A. eriantha and several species in Leiocarpae, like A. valvata, A. arguta, and A. ploygama, were highly resistant to Psa, indicating the potential presence of bacterial canker resistant genes in these Actinidia germplasm (Michelotti et al. 2018; Song et al. 2019; Wang et al. 2019; Wang et al. 2020a). A major way in plant disease resistance breeding is to introduce resistant genes to the susceptible cultivars, therefore, a survey of plant resistance gene analogs (RGAs) in the wild relatives will lay a basic foundation for improving plant resistance to disease (Tirnaz et al. 2020; Amas et al. 2023).

With the rapid development of high-throughput sequencing technologies and genomic-based methods, genomes from several *Actinidia* species and varieties have been sequenced and assembled. The first draft genome assembly of *Actinidia* (*A. chinensis* 'Hongyang') was published at 10 years ago (Huang et al. 2013). Since then, many kiwifruit genomes have been reported, including those of *A. arguta*, *A. chinensis*, *A. eriantha*, A.hemsleyana, A. latifolia, A. ploygama, A. rufa, and A. zhejiangensis (Pilkington et al. 2018; Tang et al. 2019; Wu et al. 2019; Tahir et al. 2022; Yao et al. 2022; Akagi et al. 2023; Han et al. 2023; Wang et al. 2023b; Xia et al. 2023; Yu et al. 2023; Yue et al. 2023; Lu et al. 2024;Yue et al. 2024; Zhang et al. 2024). Some of them achieved telomere-to-telomere (T2T) and gap-free level (Han et al. 2023; Wang et al. 2023b; Yue et al. 2023). Pan-genome can capture genetic diversity from different individuals or populations and reveals genomic complexity (Liu et al. 2020; Qin et al. 2021; Shang et al. 2022; Lyu et al. 2023; Yan et al. 2023). The first kiwifruit pan-genome for A. chinensis was recently generated, and a novel SV mediating fruit coloration and fruit quality was found, providing valuable information for kiwifruit genomicsassisted breeding (Wang et al. 2024). More recently, a primary gene-based pan-genome of different Actinidia speices has been constructed and facilitated the identification of a gene related to high vitamin C content (Li et al. 2024b). Although many kiwifruit genomes and resequencing data have been published, interspecific genetic diversity among Actinidia species has not been comprehensively investigated. In this study, we took advantage of 15 high quality assemblies from eight Actinidia species to build a super pan-genome to explore the genetic diversity among Actinidia. Together with resequencing data from 112 individuals of 20 species, we systematically identified SVs potentially contributed the phenotypic variations in Actinidia. Meanwhile, we investigated the interspecific relationship within Actinidia, and potential genetic architecture driving the divergence of Leiocarpae and Maculatae. Finally, we generated pan-RGA to explore the disease resistance gene resources in *Actinidia*, providing a basic understanding of resistance genes to target for genomicsbased improvement breeding.

Results

Genome assemblies and their phylogenic relationship

To construct a super pan-genome of *Actinidia* species, we collected 15 published high-quality kiwifruit genome assemblies from eight *Actinidia* species (Table S1). Among them, four assemblies were from the *Leiocarpae* group, others were from *Maculatae* group. These assemblies had total lengths ranging from 608.3 to 652.8 Mb and BUSCO scores between 93.0 and 99.3% with an average of 97.7%, indicating their high completeness. All the assemblies were annotated according to the pipeline used in our previous study (Yu et al. 2023), resulting in gene number ranging from 40,311 to 46,308 with an average BUSCO value of 94.2% for the 15 assemblies. Orthologue genes were identified and a species tree was inferred from all sets of orthogroup genes (Fig. 1a). The analyzed *Actinidia* species could be divided into two groups, i.e.

Leiocarpae group (LG) and Maculatae group (MG). LG contained A. arguta and A. ploygama, MG contained A. chinensis, A. eriantha, A. hemsleyana, A. latifolia, A. rufa and A. zhejiangensis. Molecular dating indicated that MG diverged from LG about 13.0 million years ago (Mya). The MG consisted of two separate clades, named MG1 and MG2, and they diverged from each other about 12.3 Mya (Fig. 1a). The two haplomes of A. zhejiangensis were assigned into MG1 and MG2, separately, aligned with our previous finding of the hybrid origin of A. zhejiangensis (Yu et al. 2023).

Due to the sparse kiwifruit species with genome assemblies available, the deduced phylogenic relationship can only cover limited diversity of the Actinidia genus. To address this problem, we further collected resequencing data of 114 Actinidia accessions (28 LG and 86 MG individuals) from 20 species (Table S2). Reads were mapped to the A. chinensis cv. Donghong (hereafter AcDH) genome and then used for SNP calling (Fig. 1b). Phylogenetic tree inferred from whole-genome SNPs supported the dividing of LG and MG within Actinidia, and separation of MG1 and MG2 within MG (Fig. 1b), resembling the species tree based on orthologous genes (Fig. 1a). Beside A. arguta and A. ploygama, the LG also contained A. macrosperma and A. valvata, the MG1 contained A. latifolia, A. eriantha, A. cylindrica, A. lanceolata and so on, the MG2 contained A. chinensis, A. hemsleyana, A. rufa, A. setosa, etc. The admixture analyses showed that most LG, MG1 and MG2 had distinct compositions when K=3, and a few individuals had mixed compositions. For example, the known hybrid A. zhejiangenisis had a mixed composition originated from MG1 and MG2, respectively. Principal component analysis (PCA) also suggested that these accessions could be clearly divided into three clusters except A. zhejiangenisis, which was placed between MG1 and MG2 (Fig. 1c). All these results suggested that MG diverged in two separate clades after diverging from LG.

To further detect the genetic architecture driving the divergence of LG and MG, the pairwise genetic differentiation ($F_{\rm ST}$) between the LG and MG individuals was evaluated. Regions with $F_{\rm ST}$ value above the 95th percentile were identified as highly differentiated regions (HDRs). A total of 68.08 Mb HDRs were identified between the LG and MG, and the top three chromosomes with the longest HDRs between LG and MG were Chr08, Chr23 and Chr13 (Fig. 1d, Table S3). The HDRs contained 4,334 genes, and Gene Ontology (GO) enrichment analysis showed these genes were significantly enriched in biological processes including cellular response to stress, terpenoid metabolic process and plant-type secondary cell wall biogenesis (Fig. S1). To detect select sweep in MG after divergence from LG, we scanned genomic



Fig. 1 The phylogenetic relationship and genetic variations within *Actinidia*. **a** Phylogenic tree of *Actinidia* species based on all orthologous genes. Blue dots on the tree node represent calibration times obtained from the TimeTree database (http://www.timetree.org/). **b** Maximum-likelihood phylogenetic tree and population structure of *Actinidia* individuals based on SNPs. Azh: *A. zhejiangenisis* (**c**) Principal component analysis plot showing the first two components and the percentage of variation explained by each component. **d** F_{ST} analysis across the 29 chromosomes between LG and MG. Black dotted line represent the threshold line (95th percentile). **e** XP-CLR and F_{ST} analysis of particular regions

regions with extreme allele frequency differentiation over extended linked regions using XP-CLR. A total of 43.91 Mb regions were selected in MG overlapping with 3,900 genes (Table S4), including 18 genes participated in cell wall organization or biogenesis, such as expansin gene and xyloglucan fucosyltransferase gene. Besides, a total of 8.33 Mb regions were found in both the HDR and XP-CLR selected region, overlapping 517 genes, including some trichome initiation or development genes, like *Ach-02g04420DH*, a trichome birefringence-like gene (Kabir et al. 2023), *Ach23g11790DH*, a MYB-like transcription factor homologous to *ETC1* (Wang et al. 2008), and some genes participated in cell wall organization, like *Ach-21g00200DH* and *Ach21g00210DH* encoding cellulose synthase A catalytic subunit (Fig. 1e, Table S5). Species in LG have glabrous and smooth fruit skin, while those in MG have spotted or hairy fruit skin (Huang 2016). The HDRs region and selected regions may contribute to underlie genetic basis for phenotypic variations between LG and MG.

Gene-based pan-genome of Actinidia

The 15 kiwifruit assemblies included four LG, four MG1 and seven MG2 assemblies and were obtained five male and nine female individuals (Table S1). Ortholog investigation classified all genes into 61,465 families comprising 14,492 core (present in all 15 assemblies), 5,347 softcore (present in all 13-14 assemblies, over 85% assemblies), 21,326 dispensable (present in all 2–12 assemblies), and 20,300 cloud (present in only one assembly) gene families (Fig. 2a). Although dispensable and cloud gene families accounted for a larger proportion (67.72%) of the pan-genome, as the average numbers of genes belonging them were much less than those of core and softcore gene families (Fig. S2), they occupied less than a quarter of total genes. Each assembly had an average of 3.37% cloud genes (Fig. 2a, 2b). Meanwhile, core gene families contained over half of the total genes, and at least 54.09% for each assembly.

We found that 91.10% of the core genes, and 85.07% of the softcore genes contained InterPro domains, which were much higher than the percentages in the dispensable and cloud genes (55.37% and 34.57%, respectively) (Fig. 2c). Besides, the average length and exon number of core genes were significantly longer than other groups of genes (Fig. 2d, 2e), indicating that gene structure was more complex in core genes than softcore and dispensable genes. Meanwhile, core genes had lower nucleotide diversity (π) in their coding sequences than softcore genes, but higher than the dispensable genes (Fig. 2f), suggesting that the core genes were more functionally conserved than softcore genes. Additionally, core genes had both higher percentage of expressed genes (fragments per kilobase of exon model per million mapped fragments (FPKM)>0.5) and higher expression levels than softcore and dispensable genes in different tissues (Fig. 2g, Fig. S3), GO enrichment analysis showed that core genes were enriched for several essential biological processes, including nitrogen compound and phosphorus metabolic processes, RNA processing, chromosome organization and cellular response to DNA damage stimulus (Fig. S4), while the softcore and dispensable genes were enriched in response to stress, defense response, response to oxidative stress, and cell wall organization or biogenesis etc. (Fig. 2h).

A total of 74 gene families (containing 361 genes) were found in each LG but not in any MG genome, that is, specific to LG, including several gene families involved in cell wall organization, like genes encoding trichome birefringence-like protein or glycoside hydrolase (Table S6). These specific gene families probably contribute to the fruit characteristics, like smooth and hairless fruit skins of LG. Meanwhile, 140 gene families (containing 608 genes) were specific to MG1, including gene families encoding redox enzymes, like cytochrome P450, NADdependent epimerase and ferredoxin reductase, and one gene family encoding hydroxyacylglutathione hydrolase, which plays vital role in glutathione (GSH) metabolism (Dorion et al. 2021). Previous study reported that GSH participated in the AsA metabolism in A. eriantha via AsA-GSH cycle (Liao et al. 2021), thus this MG1-specific gene families might play a role in the AsA metabolism in A. latifolia and A. eriantha, which had high vitamin C content in fruits. Besides, gene families specific to male accessions included a gene family homologous to FrBy, a known Actinidia sex-determining genes (Akagi et al. 2019), and two gene families with unknown function, which didn't not locate in the known Actinidia sex-determining region (Table S7), suggesting this gene family was probably not related to sex differentiation. Meanwhile, no gene families specific to female accessions were found.

Global landscape of structural variations in Actinidia

To identify SVs, both genome assemblies and thirdgeneration sequencing (TGS) data of 11 accessions were aligned against the AcDH genome by PanPop (Table S8), a tool that can enhanced SV accuracy by merging and filtering of SVs from multiple SV callers (Zheng et al. 2024). As for A. zhejiangenisis and another two Actinidia accessions without public assemblies (A. chinensis 'H0809' and A. eriantha 'Blank'), SVs were investigated using only TGS data. Thus, SVs in 14 Actinidia accessions from eight species were identified. In short, except A. zhejiangenisis, we identified an average of 164,777 SVs $(\geq 50 \text{ bp in size})$ per accession (ranging from 77,536 SVs for A. chinensis 'Hongyang' to 228,098 for A. eriantha), affecting an average of 96.54 Mb of sequence per accession (ranging from 63.16 Mb for A. chinensis 'H0809' to 125.44 Mb for A. hemsleyana) (Fig. 3a). A. zhejiangenisis, which contained genetic information of two parent species resulting the largest number of SVs (270,419). Insertions and deletions accounted for the majority of SVs, and inversions were much less than insertions and deletions. If excluding A. zhejiangenisis, no significant difference in SV numbers or total lengths were found between LG and MG1, or MG1 and MG2 accessions. Most SVs located in intergenic and intronic regions, then upstream and downstream of genes. All the SVs were merged to generate a graph-based genome with a set of 1,277,140 non-redundant variations.

We further mapped the resequencing short reads of the 112 *Actinidia* accessions (Table S2) onto the graphbased pan-genome. After filtering, a total of 162,866 variations were obtained, including 80,803 insertions and 82,063 deletions. The NJ tree based on the SVs showed these individuals could be classified into three groups: LG, MG1 and MG2, and similar result was found in PCA





Fig. 2 Gene-based pan-genome of *Actinidia*. (a) Composition of the pan-genome. The histogram shows the number of gene families in the 15 assemblies with different frequencies. Pie shows the proportion of core, softcore, dispensable, and cloud genes. **b** The core, softcore, dispensable, and cloud gene numbers in each assembly. **c** Proportions of genes with and without InterPro domains in core, soft core, dispensable, and cloud genes. The distribution of gene lengths (**d**), exon numbers (**e**) and nucleotide diversity (**f**) of core, softcore and dispensable genes. ***p < 0.001, Student's *t* test. **g** Expression levels of core, softcore and dispensable genes in *A. eriantha* at different tissues. **h** GO enrichment analysis of softcore and dispensable genes

analysis (Fig. S5) in line with the SNP analyses (Fig. 1c). Besides, we found an uneven distribution of SVs along the chromosomes, and 532 SV hotspot regions were

identified (Fig. 3b, Table S9). The top three chromosomes with longest SV hotspot regions were Chr15 (4.16 Mb), Chr20 (4.12 Mb), Chr08 and Chr03 (4.02 Mb), that is, the



Fig. 3 Structural variations and graph-based pan-genome of *Actinidia*. **a** The SV numbers, lengths and percentage of SVs in indicated regions in each accession. **b** SV numbers on chromosomes in 200 kb windows. Regions over the threshold line (95th percentile) are SV-hotspot regions. **c** The percentage of DEGs in SV-hotspot genes, SV-hotspot-upstream genes and non-SV-hotspot genes under copper, heat and waterlogging stresses. **d** SV frequency of LG-, MG1-, and MG2-favored SVs. The broken lines indicate the averaged frequencies, and the ribbons indicate the ranges of SV frequencies. **e** Genotypes of three SVs in LG and MG groups. (**f**) Expression levels of *Ach13g13590DH* and *Ach25g04680DH* in the fruit peels from three kiwifruit species. ***p < 0.001, *p < 0.05, Student's t test

most dynamic chromosomes. GO enrichment analysis showed that the genes in SV hotspot regions were significantly enriched in plant organ formation, which might be associated with the phenotypic variations among different *Actinidia* species, and regulation of gene expression (Fig S6). We therefore explored the general expression characterization of SV-hotspot genes (genes with gene sequences overlapping with SV-hotspot regions),

SV-hotspot-upstream genes (genes with 2 kb upstream overlapping with SV-hotspot regions) and non-SVhotspot genes (neither SV-hotspot, nor SV-hotspotupstream genes) using RNA-seq data of A. chinensis under different abiotic stress conditions including copper, heat and waterlogging stresses (Fig. 3c). The percentage of significantly differentially expressed genes (DEGs) were calculated in the three types of genes, and we found that both the SV-hotspot and SV-hotspot-upstream genes had higher percentages of DEGs than non-SVhotspot genes. Moreover, the SV-hotspot or SV-hotspotupstream genes had obviously higher proportion of genes with > fourfold expression changes after stress treatment than the proportion of non-SV-hotspot genes, indicating that SV-hotspot and SV-hotspot-upstream are generally more sensitive to environmental stresses. These results suggesting SV hotspot regions may undergo stronger environmental selection compared to other genome regions, which was consistent with the previous findings in rice and *Malus* accessions (Qin et al. 2021; Wang et al. 2023a).

The frequency of each SV in LG, MG1 and MG2 individuals except A. zhejiangenisis was calculated, respectively, and LG-favored SVs with high frequency (>0.5) in LG but low frequency (< 0.05) in MG1 and MG2 individuals, were identified, as well as MG1- and MG2-favored SVs. A total of 18,364 LG-favored, 995 MG1-favored and 3,743 MG2-favored SVs were identified (Fig. 3d). Among the LG-favored SVs, nearly three quarters of SVs only existed in LG individuals, and these SVs overlapping with the genes enriched in various biological processes, such as nitrogen compound metabolic process, phosphorus metabolic process and cellular response to stimulus (Fig S7). It's worth noting that some of them were related to trichome development. For example, a 55-bp insertion in the exon of Ach22g02880DH, a gene homologous to AtPIR gene, which played role in actin filament reorganization and trichome development (Isner et al. 2017); a 114-bp deletion in the second intron of *Ach19g03580DH*, a gene homologues to WER, which encodes a MYB transcript factor regulating non-hair cell fate (Lee et al. 1999), a 114-bp insertion in the last exon of Ach16g02920DH, a gene encoding polygalacturonase, were found present in most LG individuals, but absent in MG (Fig. 3e, Table S10). Besides, 4,784 LG-favored SVs located in the promoter regions of 5,230 genes. For instance, a 218-bp insertion on the promoter of Ach13g13590DH, a gene homologous to CPC, and a 205-bp insertion on the promoter of Ach25g04680DH, a gene homologous to GL3. Both CPC and GL3 were major regulators of hair cell fate establishment (Wang et al. 2022). We then compared the expression levels of Ach13g13590DH and Ach25g04680DH in the fruit peels from three kiwifruit species, A. arguta in LG with hairless fruit skin, A. eriantha in MG1 hairy fruit skin, and A. indochinensis in MG2 with spotted fruit skins. Their expression levels in A. arguta were significantly lower than that in A. eriantha or A. indochinensis (Fig. 3f), suggesting that these two genes might contributed to the diversity of fruit skin of Actinidia.

The MG1-favored and MG2-favored SVs overlapped with 263 and 1,431 genes, respectively. These genes participated in various biological processes, including establishment of localization, cellular response to stimulus, regulation of biological process, cell wall organization or biogenesis and so on (Table S11). Besides, we found 21.48% of all the favored SVs located in the SV-hotspot regions. Overall, those SVs favored or specific in given group might contributed to the phenotypic diversity or environmental adaptation of *Actinidia*.

Disease resistance genes reservoir and Pan-RGA of Actinidia

RGAs are potential *R* genes with specific conserved domains and motifs. To explore the reservoir of disease resistance genes in *Actinidia* species, we identified all classes of RGAs from the 15 *Actinidia* assemblies (Table S1). In total, 18,858 RGAs were detected, including 2,626 nucleotide-binding site (NBS)-encoding proteins, 11,495 receptor-like protein kinases (RLKs), 1,624 receptor-like proteins (RLPs) and 3,113 proteins with both transmembrane domain and coiled-coil domain (TM-CCs) (Fig. 4a). The identified RGAs in each assembly ranging from 1,110 to 1,494, and the difference in RGA numbers was not only observed between species, but also within species. RLKs accounted for the largest portion of RGAs in each assembly, and almost randomly distributed on the 29 chromosomes (Fig. S8). Meanwhile,

(See figure on next page.)

Fig. 4 Resistance gene analogs (RGAs) in the 15 Actinidia assemblies. **a** RGA gene numbers in each assembly. **b** NBS gene number on each chromosome. **c** The percentage of genes located in clusters of each type. **d** A representative NBS cluster on chromosome 19 has conserved paired NLRs among the Actinidia accessions, and seven representative accessions are used here for displaying. **e** Expression levels of genes in the NBS cluster after *Psa* infection. Three representative accessions are used here. **f** Composition of the pan-RGA. The histogram shows the number of gene families in the 15 assemblies with different frequencies. Pie shows the proportion of core, softcore, dispensable, and cloud RGA genes. **g** Percentages of core, softcore, dispensable, and cloud genes in each type of RGAs



Fig. 4 (See legend on previous page.)

NBS genes exhibited very uneven distribution pattern on chromosomes comparing to other types of RGAs, with Chr08, Chr10 and Chr20 containing significantly higher numbers of NBS genes than other chromosomes (Fig. 4b). These RGAs, combined with molecular markers from known disease resistance QTL, could be used to identify candidate resistance genes. For example, we found some RLKs in the previous identified QTL regions for *Psa* resistance on Chr27 (Tahir et al. 2019). Among them, one RLK gene, *Ach27g03710DH*, displaying presence/absence variations (PAVs) among different species, might contribution to response to *Psa* infection.

The genes belonging to same family and within 200 kb of each other in the genome were taken as gene clusters (Van de Weyer et al. 2019). We found most NBS genes in each assembly tend to locate in such clusters, and the percentage of clustered genes in NBS genes was significantly higher than those in RLKs, RLPs and TM-CCs (Fig. 4c). The NBS gene cluster mainly constituted of 2–5 genes, and the largest cluster in each assembly constituted of 6-15 genes (Fig. S9). Paired nucleotide-binding leucinerich repeat (NLR) genes, a particularly interesting subset of NBS genes in head-to-head orientation (van Wersch and Li 2019), were identified. Each accession had an average of 6.77% paired NLRs of NBS genes, and we found one pair of paired NLR genes evolutionarily conserved in all analyzed Actinidia accessions (Fig. 4d). In AcDH, this gene pair, Ach19g11900DH and Ach19g11910DH, located in a NBS cluster of five genes on chromosome 19. We further explored the expression levels of genes in this NBS cluster after Psa infection in AcDH, A. eriantha 'Midao31' and A. arguta 'M1'. The genes in this cluster exhibited inconsistent expression patterns at three stages, whereas the two genes of the paired NLR showed similar expression levels in all stages (Fig. 4e), suggesting that this paired NLR probably not only evolutionarily conserved, but also functionally linked for Actinidia defense response.

Integrated domains (IDs) of NBS genes, which participate in interaction with effectors, were identified. The top three most frequent IDs of NBS were AAA ATPase, C-JID and NACHT domains (Fig. S10), which were shared by all accessions. Meanwhile, different accessions exhibited different IDs compositions, for instance, only the three conserved ID domains were shared by the three *A. arguta* accessions. Additionally, 17, 22 and 28 ID domains were found in different *A. chinensis, A. eriantha* and *A. arguta* accessions, respectively (Fig. S11). Besides, 10, 15 and 18 IDs were unique to LG, MG1 and MG2, respectively. The diversity of ID domain distribution within and among *Actinidia* species may be associated to their different performance of disease resistance.

Multiple mutants were found on the RGA genes, indicating the abundant genetic diversity of RGAs in Actinidia. We further built a pan-RGA of Actinidia species to reveal the diversity of disease resistance genes. These RGAs were grouped into 1,426 gene families based on sequence similarity, including 263 core (present in all 15 assemblies), 223 softcore (present in all 13-14 assemblies), 623 (present in all 2–12 assemblies) and 317 cloud gene families (present in only one assembly) (Fig. 4f). At gene level, the pan-RGA comprised 7,893 core, 5,486 softcore and 5,136 dispensable genes, accounted for 41.85%, 29.09%, and 27.23% of all RGAs, respectively. Comparing to the whole pan-genome, pan-RGA had much lower percentage of core genes, but more percentage of softcore genes. Additionally, RLK genes had higher percentages of core genes than other type genes (Fig. 4g). For each assembly, core, softcore and dispensable genes had a slight difference in number (Fig. S12). Besides, we found 184 dispensable gene families (575 genes) absent in A. chinensis, the most commercialized kiwifruit species, and 28 dispensable gene families (67 genes) absent in MG (Table S12). These genes may be valuable resource for future resistance breeding of kiwifruit.

Discussion

Actinidia was previously proposed to be divided into Leiocarpae and Maculatae group based on morphological evidences (Li et al. 2000), while the further classification of Maculatae was not sustainable (Huang 2016). Classification based on morphological evidence has some limitations as some species exhibit abundant morphological variations. Therefore, here we explored the evolutionary relationships within Actinidia by genomic approaches. We generated phylogenic trees based on ortholog genes, SNPs and SVs (Fig. 1 and Fig. S5, S14), and all the results showed that Maculatae group could be subdivided into two subgroups. Huang et al. (1999) suggested dividing Actinidia into three sections: Leiocarpae, members in this section have smooth-skinned fruits; Maculatae includes the species with spotted fruits, and Vestitae has leaf hairs base. Nevertheless, the MG1 and MG2 groups did not correspond to Maculatae and Vestitae groups suggested by Huang et al. (1999). Moreover, no common distinguishable phenotypic difference between MG1 and MG2 was observed in terms of the reported 50 morphological characters of Actinidia (Huang et al. 1999), which may be due to ongoing interspecific hybridization and gene flow between MG1 and MG2 (Liu et al. 2017; Yu et al. 2023). Yet we could not rule out the presence of the distinguishable phenotypes between MG1 and MG2 beyond the scope of previous studies.

Besides, we noticed that the phylogenic tree based on single copy gene had a minor difference at the topology of MG2 with the trees based on all orthologs. Specifically speaking, *A. chinensis* var. *deliciosa* was outside of the *A. chinensis* var. *chinensis* and *A. rufa* branches in the tree based on single copy genes (Fig S13), while closely related to *A. chinensis* var. *chinensis* in the tree based on all orthologs (Fig. 1a). This could be caused by the interspecific gene flow among the MG2, which was mentioned in other study (Liu et al. 2017), or the incomplete lineage sorting (ILS) of ancestral polymorphisms. Moreover, the *A. chinensis* var. *deliciosa* individuals did have mixed component in the admixture analysis when K=3, suggested potential interspecies hybridization between *A. chinensis* and other *Actinidia* species.

Fruit hair is an important appearance quality affecting the market value of kiwifruits. Leiocarpae group has smooth-skinned, hairless fruit, such as A. arguta with edible and hairless fruit skin; while *Maculatae* group has spotted or hairy fruit skin, such A. chinensis var. deliciosa with rough and hairy skins, which is not convenient for direct consumption. In this study, we tried to uncover the genetic basis for phenotypic variations of fruit skin between Leiocarpae and Maculatae groups using different approaches. In short, we found a number of genes associated with trichome initiation or development and plant-type cell wall organization might contribute to the fruit skins variations in Actinidia, like trichome birefringence-like genes, some genes for MYB transcription factors, pectinesterase or polygalacturonase and so on. Trichome development can be divided into six stages, and the last stage of its development is maturation of the cell wall (Kubátová et al. 2019). The secondary cell wall consists of the outer, cellulose-rich layer and the inner, callose-rich layer for Arabidopsis trichomes (Kulich et al. 2015). As genes related to trichome development and plant-type cell wall organization were found associated with the phenotypic diversity in Actinidia, Thus, we speculated that *Actinidia* species might have diverse secondary cell wall components of trichomes. On the other hand, it was found that A. arguta fruits soften quickly and show a much shorter shelf life than A. chinensis, which might due to the different expression levels of CEL1, a gene encoding cellulose, and PME1, a gene encoding pectin methylesterase (Lu et al. 2024). Therefore, the differentiated cell wall biogenesis or organization related genes might not only have roles in trichome biogenesis and development, but also affect fruit ripening or shelf life between *Leiocarpae* and *Maculatae* groups.

Increasing evidence showed that SVs are responsible for many phenotypic variations (Qin et al. 2021; Lyu et al. 2023; Yan et al. 2023; Wang et al. 2023a). Here, we generated a graph-based pan-genome integrating SVs from 15 Actinidia accessions, and further identified SVs based on the pan-genome at population scale (Fig. 3). Our result showed SVs can also support inferences about the evolutionary relationships as SNPs. Moreover, SV hotspot regions, the most dynamic genomic region in Actinidia, contributed to responses to environmental pressures, as SV-hotspot and SV-hotspot-upstream genes were more sensitive to abiotic stresses than genes in other regions. Actinidia has a wide distribution in eastern Asia, and different groups of Actinidia taxa had characteristic geographic distributions, e.g., the Leiocarpae were to be found mainly in north China with relative cold and dry condition (Huang et al. 2016). Thus, the dynamic genomic variations might contribute to environmental adaptation of Actinidia (Zhang et al. 2023). Besides, some SVs were found selected in particular group, and many LG-favored SVs altered gene sequences or regulatory sequences of genes participating in trichome development or cell wall organization. Therefore, SVs might regulate the phenotypic diversity of Actinidia and drive the differentiation of MG and LG by changing gene structure or expression of nearby genes. Our study analyzed a small Actinidia population, and further studies should concentrate on SVs in larger Actinidia population along with accurate phenotypic or environmental data, which will allow SV-based association studies for trait associated gene discovery and improvement breeding.

The sustainable control of plant pathogens relies on the application of genetic resistance primarily driven by RGAs. Under the guidance of the gene-based pangenome, we systematically identified all the RGAs across the 15 Actinidia assemblies and have generated a pan-RGA dataset for Actinidia (Fig. 4). Although some studies had identified disease genes in A. chinensis and A. eriantha (Wang et al. 2020b; Yao et al. 2022), we firstly revealed RGA diversity in Actinidia on a large scale covering the LG and MG. We found most NBS genes in clusters, and one paired NLR evolutionarily conserved in Actinidia, exhibited similar expression levels after Psa infection. Besides, a total of 205 gene families from pan-RGA were found absent in all A. chinensis assemblies. These genes greatly enriched the disease resistance gene reservoir for future resistance breeding of kiwifruits. The pan-RGA dataset could not only reveals the disease gene diversity in Actinidia, but also be used to identify candidate resistant genes combined with molecular markers from known disease resistance QTL.

In conclusion, the pan-genome construction and genetic variation identification in this study comprehensively reveal the genetic diversity of *Actinidia* species, which will greatly benefit the *Actinidia* breeding and functional genomics research.

Methods

Genome sequences collection and annotation

Previously published 15 genome assemblies of eight Actinidia species, including A. arguta, A. chinensis, A. eriantha, A.hemsleyana, A. latifolia, A. ploygama, A. rufa, and A. zhejiangensis, as well as an outgroup, R. ovatum genome were downloaded from China National Genomics Data Center, Figshare database, GitHub and National Center for Biotechnology Information (NCBI) (Table S1). As A. zhejiangensis was demonstrated as F_1 hybrid of A. eriantha and A.hemsleyana in previous study (Yu et al. 2023), the two haplomes with different origins were taken as two assemblies in this study. For other haplotype-resolved genomes, the haplome with the longest sequence or most gene number was chosen as representative (Table S1). BUSCO analysis (v5.0) was performed with the 'eudicots odb10' database to evaluate genome/ gene completeness (Manni et al. 2021). All the other assemblies were aligned to the T2T genome of AcDH to check the chromosome order and direction by MUMmer v4.0.0rc1 program (Marcais et al. 2018).

To ensure the quality of gene-based pan-genome analysis, all the assemblies were de novo annotated by the same pipeline according to those used for A. hemsleyana and A. zhejiangenisis in our previous study (Yu et al. 2023), which incorporated transcriptome, ab initio and homolog predictions were used predict protein-coding genes. RNA-seq reads were downloaded from public database (Table S13). Adapter sequences and low-quality reads were removed by fastp v0.22.0 (Chen et al. 2018). Clean reads were then aligned to each assembly using HISAT2 v2.2.1 (Kim et al. 2019), and assembled using StringTie2 v2.1.6 (Kovaka et al. 2019). The coding regions were predicted using TransDecoder v5.5.0 (http://trans decoder.github.io). For ab initio prediction, gene models were predicted using AUGUSTUS v3.3.3 (Stanke et al. 2008) and GlimmerHMM (Majoros et al. 2004). Protein sequences of Solanum lycopersicum, Arabidopsis thaliana, A. zhejiangenisis and Oryza sativa were used for homolog-based predictions. Finally, the information of repetitive sequences, transcripts, ab initio and homologybased gene predictions were integrated using MAKER v2.31.11 (Cantarel et al. 2008).

Gene family clusters and phylogenetic analysis

OrthoFinder v2.5.2 (Emms and Kelly 2019) was used to infer a matrix of orthologous groups (gene families) among all the *Actinidia* genomes and *R. ovatum* genome. A species tree was inferred from all sets of orthogroup genes including single-copy and multi-copy orthogroups using STAG (Emms and Kelly 2018). Besides, multiple sequence alignments were performed using MUS-CLE v3.8.31 (Edgar 2004) for the identified single-copy orthologous genes. A maximum likelihood (ML) phylogenetic tree was constructed using the alignments of singlecopy orthologous genes with RAxML v8.2.12 (Stamatakis 2014). Species divergence time estimates were calculated using MCMCTREE in PAML (v4.9i) (Yang 2007), and calibrated using the estimated divergence times for *R. ovatum* and *A. chinensis* (53.4–99.2 Mya), *A. arguta* and *A. chinensis* (5.0–35.0 Mya) in the TimeTree database (http://www.timetree.org/). Gene family expansion or contraction was determined using CAFÉ (v3.0) (De Bie et al. 2006).

SNP calling and phylogenetic analysis

Genome resequencing data of 114 Actinidia accessions from 20 species were downloaded from the NCBI Sequence Read Archive database (SRA) and Genome Sequence Archive (GSA) database (Table S2). Adapter sequences and low-quality reads were removed by fastp v0.22.0 (Chen et al. 2018). The clean reads were then mapped to the AcDH genome using BWA v0.7.17, and duplicated reads were further removed by MarkDuplicates of GATK v4.2.0 (DePristo et al. 2011). SNP calling was performed using GATK HaplotypeCaller, generating a single variant calling file (VCF). The VCF was preliminarily filtered by GATK VariantFiltration based on following criteria: QD < 2.0 || MQ < 40.0 || FS > 60.0 || SOR>3.0 || MQRankSum<12.5 || ReadPosRank-Sum < 8.0, and further filtered by VCFtools v0.1.16 (Danecek et al. 2011) with the following parameters: maf 0.1 -minQ 30 -mac 3 -minDP 3 -max-missing 0.5.

PCA was performed by PLINK v1.90 (Purcell et al. 2007). The ancestral population structure was estimated using ADMIX TURE v1.3.0 (Alexander et al. 2009) with number of sub-populations (*K*) from 2 to 5. For phylogenetic relationship analysis, SNPs were thinned using a distance filter of interval \geq 2000 bp. A ML tree was constructed to investigate genetic relationships within *Acitinida* by IQ-TREE (Minh et al. 2020) with 1000 rapid bootstraps. To detect selective sweeps, SNPs were subjected to XP-CLR (Chen et al. 2010) with 50-kb sliding window and 20-kb step for each chromosome. The top 5% XP-CLR values across the genome were considered be potential selected loci. $F_{\rm ST}$ was calculated in 20 kb stepping windows using VCFtools.

Gene-based pan-genome construction and analysis

We constructed a gene-based pan-genome using the 15 *Actinidia* assemblies (Table S1). The core, softcore, dispensable and cloud gene sets among the 15 genomes were estimated based on gene family clustering using OrthoFinder v2.5.2 with an inflation parameter of 1.5. For each gene family, one gene was randomly chosen, and the representative protein sequences of all gene

families were aligned against InterPro databases with Diamond v2.0.13.151 (Buchfink et al. 2021). GO terms were assigned according to the InterPro classification. GO enrichment analysis with gene families was performed using Fisher's exact test with an adjusted *P*-value (q value) of < 0.05.

To analyze the length and exon numbers of core, softcore and dispensable genes, all core, softcore and dispensable genes of A. arguta, A. eriantha and AcDH were used. Nucleotide diversity (π) was calculated by VCFtools. The RNA-seq data used for calculated the expression levels of core, softcore and dispensable genes were obtained from public databases (Table S13). Raw reads were filtered to remove adapter and low-quality sequences by fastp. Cleaned reads were mapped to the corresponding assembly using HISAT2 v2.2.1 (Kim et al. 2019), and gene expression levels were calculated using StringTie v2.1.6 (Kovaka et al. 2019). To find the gene families related to sex determination, gene families present in each male assembly but absent in any female assembly, that is, specific to male accessions were analyzed.

SVs identification and graph-based *pan*-genome construction

TGS data was downloaded from public databases (Table S8). TGS data and genome assemblies were used to identify SVs against AcDH using PanPop (Zheng et al. 2024), a sequence-aware SV merging and processing pipeline, with default parameters. For A. zhejiangenisis, which contained genome information of two species, only TGS data was used for SV calling. Besides, SVs were investigated using only TGS data for another two Actinidia accessions with public TGS data but not assemblies (A. chinensis 'H0809' and A. eriantha 'Blank'). To enhance accuracy of SVs, five SV callers, sniffles, cuteSV, svim, pbsv, and Assemblytics were used, and SVs supported by at least two callers were retained. The merged nonredundant variations in VCF format were used to construct the graph-based pan-genome by vg toolkit v1.50.1 (Garrison et al. 2018).

To identify population-scale SVs, the clean short reads of the 112 accessions (Table S2, two individuals (*A. chinensis*_5 and *A. chinensis*_6) with single-end sequencing data were removed here) were mapped to the graphbased genome by giraffe and SVs were then filtered and merged using PanPop with default parameters. Those SVs were further filtered by PLINK with the following parameters: –geno 0.5 –maf 0.05. PCA analysis was performed via PLINK. To build a phylogenic tree with SVs, *P*-distance matrix based VCF was generated by VCF2Dis (https://github.com/BGI-shenzhen/VCF2Dis), and then used to construct a neighbor-joining tree by FastME 2.0 (Lefort et al. 2015). To identify SV hotspot regions, we calculated the distribution of SVs for each 100 kb window (with a 20 kb step size) along each chromosome. The top 5% of all windows with the highest frequency of variations were selected, and then merged as SV hotspot regions. SVs in intergenic, upstream, downstream, intron and exon regions were annotated using ANNO-VAR (Wang et al. 2010), and the number was calculated in each accession. To analyze the frequencies of SVs, the genotyped SVs of each group were obtained by VCFtools, and the frequencies of SVs were then calculated, respectively. RNA-seq data used here was obtained from public databases (Table S13). Gene expression levels were calculated by StringTie, and read counts were estimated by prepDE.py script of StringTie. Differentially expressed genes were identified by DESeq2 (Love et al. 2014).

RGA identification and pan-RGA analysis

The deduced protein sequences of the annotated genes of all 15 assemblies were used to identify RGAs via the RGAugury pipeline (Li et al. 2016). The identified RGA candidates included NBS-encoding proteins, RLKs, RLPs and TM-CCs. The syntenic gene pairs between species were identified using JCVI v.1.3.8, a Python version of MCscan (Tang et al. 2008). IDs of NBS proteins were predicted using HMMER 3.1b2 (https://github.com/EddyR ivasLab/hmmer). To analyze the expression pattern of paired NLR or clustered NBS genes in disease response, RNA-seq data of A. chinensis, A. eriantha and A. arguta under Psa treatment was downloaded (Table S13), and clean reads were mapped to AcDH, A. eriantha 'Midao31' and A. arguta 'M1' using HISAT2, respectively. Gene expression levels were then calculated by StringTie. Gene family clustering was performed with all RGAs by OrthoFinder and core, softcore, dispensable and cloud RGA gene sets were subsequently estimated.

Abbreviations

A. arguta	Actinidia arguta
A. chinensis	Actinidia chinensis
A. eriantha	Actinidia eriantha
A. hemsleyana	Actinidia hemsleyana
A. latifolia	Actinidia latifolia
A. ploygama	Actinidia polygama
A. rufa	Actinidia rufa
A. valvata	Actinidia valvata
A. zhejiangensis	Actinidia zhejiangensis
AcDH	A. chinensis cv. Donghong
AsA	Ascorbic acid
DEG	Differentially expressed gene
FPKM	Fragments per kilobase of exon model per million mapped
	fragments
FST	Fixation index
GO	Gene ontology
GSA	Genome Sequence Archive
GSH	Glutathione
HDR	Highly differentiated region
ID	Integrated domain

ILS	Incomplete lineage sorting
LG	Leiocarpae group
MG	Maculatae group
ML	Maximum likelihood
Mya	Million years ago
NAP1	Nck-associated protein 1
NCBI	National Center for Biotechnology Information
NBS	Nucleotide-binding site
NLR	Nucleotide-binding leucine-rich repeat
PAV	Presence/absence variant
Psa	Pseudomonas syringae pv. Actinidiae
PCA	Principal component analysis
QTL	Quantitative trait locus
RGA	Resistance gene analog
RLK	Receptor-like protein kinase
RLP	Receptor-like protein
SNP	Single nucleotide polymorphism
SRA	Sequence Read Archive database
SV	Structural variation
T2T	Telomere-to-telomere
TM-CC	Protein with both transmembrane domain and coiled-coil domain
TGS	Third-generation sequencing
VCF	Variant calling file π Nucleotide diversity

Supplementary Information

The online version contains supplementary material available at https://doi. org/10.1186/s43897-024-00123-1.

Additional file 1: Fig. S1. GO enrichment analysis of genes in the highly differentiated regions between LG and MG by F_{ST} analysis. Fig. S2. Gene number in each gene family of the pan-genome. Fig. S3. Expression levels of core, softcore and dispensable genes in different species. (a) Expression levels of genes in the fruits of A. chinensis 'Donghong' at different days after flowering (DAF). (b) Expression levels of genes in different tissues from A. arguta 'M1'. Fig. S4. GO enrichment analysis of core genes from the gene-based kiwifruit pan-genome. Fig. S5. Phylogenic tree (a) and principal component analysis (b) based on SVs from 112 Actinidia accessions. Azh: A. zhejiangensis. Fig. S6. GO enrichment analysis of genes in SV hotspot regions. Fig. S7. GO enrichment analysis of genes overlapping with SVs with high frequency (> 0.5) in LG but absent in MG. Fig. S8. Number of RLK, RLP and TM-CC genes on each chromosome. Fig. S9. NBS gene cluster size and frequency in each assembly. Fig. S10. Word cloud of integrated domains of NBS genes in Actinidia. Fig. S11. Venn diagram of NBS integrated domains of selected species. Fig. S12. Proportion of core, softcore, dispensable and cloud gene family from pan-RGA in each assembly. Fig. S13. Phylogenic tree of Actinidia species based on singlecopy orthologous genes.

Additional file 2: Table S1. All the *Actinidia* assemblies used in this study. Table S2. The genome resequencing data used in this study. Table S3. Highly differentiated regions between LG and MG. Table S4. Selected regions of MG by XP-CLR analysis. Table S5. Genes in both the HDR and XP-CLR selected region. Table S6. Gene families specific to LG or MG. Table S7. Gene families specific to male accessions. Table S8. The genome assemblies and third-generation sequencing data used for graph-based pan-genome construction. Table S9. SV hotspot regions in *Actinidia*. Table S10. Details of SVs involved in this study. Table S11. The genes overlapping with MG1- or MG2-favored SVs. Table S12. The RGA gene families absent in *A. chinensis* and MG. Table S13. RNA-seq data used in this study.

Acknowledgements

We thank Prof. Dawei Li (Wuhan Botanical Garden, Chinese Academy of Sciences) for his valuable comments and suggestions.

Authors' contributions

LG, XYao and SG designed the research. XYu analyzed data and wrote the manuscript. PW, SG and SL participated in data collection. MQ, MZ, EL and HL participated in data analysis. LG and XYao revised the manuscript. All authors read and approved the final manuscript.

Funding

This work was supported by grants from the Hubei Provincial Natural Science Foundation of China (2024AFA035), the National Natural Science Foundation of China (32170395 and 32070377), and the Foundation of Hubei Hongshan Laboratory (2021hszd017).

Data availability

The pan-genome datasets and all the identified RGAs in *Actinidia* can be found at https://figshare.com/articles/dataset/KiPan/26075293.

Declarations

Ethics approval and consent to participate

Not applicable.

Consent for publication

All authors approve the manuscript and consent to publication of the work.

Competing interests

The authors declare that they have no competing interests. The authors declare that they have no financial or nonfinancial interests.

Received: 7 January 2024 Accepted: 24 October 2024 Published: 24 January 2025

References

- Akagi T, Pilkington SM, Varkonyi-Gasic E, Henry IM, Sugano SS, Sonoda M, et al. Two Y-chromosome-encoded genes determine sex in kiwifruit. Nat Plants. 2019;8:801–9.
- Akagi T, Varkonyi-Gasic E, Shirasawa K, Catanach A, Henry IM, Mertten D, et al. Recurrent neo-sex chromosome evolution in kiwifruit. Nat Plants. 2023;3:393–402.
- Alexander DH, Novembre J, Lange K. Fast model-based estimation of ancestry in unrelated individuals. Genome Res. 2009;19(9):1655–64.
- Amas JC, Bayer PE, Hong Tan W, Tirnaz S, Thomas WJW, Edwards D, et al. Comparative pangenome analyses provide insights into the evolution of *Brassica rapa* resistance gene analogues (RGAs). Plant Biotechnol J. 2023;1(10):2100–12.
- Buchfink B, Reuter K, Drost H-G. Sensitive protein alignments at tree-of-life scale using DIAMOND. Nat Methods. 2021;8(4):366–8.
- Cantarel BL, Korf I, Robb SM, Parra G, Ross E, Moore B, et al. MAKER: an easyto-use annotation pipeline designed for emerging model organism genomes. Genome Res. 2008;18(1):188–96.

Chen H, Patterson N, Reich D. Population differentiation as a test for selective sweeps. Genome Res. 2010;20(3):393–402.

- Chen S, Zhou Y, Chen Y, Gu J. Fastp: an ultra-fast all-in-one FASTQ preprocessor. Bioinformatics. 2018;4(17):1884–90.
- Danecek P, Auton A, Abecasis G, Albers CA, Banks E, DePristo MA, et al. The variant call format and VCFtools. Bioinformatics. 2011;7(15):2156–8.
- De Bie T, Cristianini N, Demuth JP, Hahn MW. CAFE: a computational tool for the study of gene family evolution. Bioinformatics. 2006;22(10):1269–71.

DePristo MA, Banks E, Poplin R, Garimella KV, Maguire JR, Hartl C, et al. A framework for variation discovery and genotyping using next-generation DNA sequencing data. Nat Genet. 2011;43(5):491–8.

Dorion S, Ouellet JC, Rivoal J. Glutathione metabolism in plants under stress: Beyond reactive oxygen species detoxification. Metabolites. 2021;11(9):641. Edgar RC. MUSCLE: multiple sequence alignment with high accuracy and high throughput. Nucleic Acids Res. 2004;32(5):1792–7.

Emms DM, Kelly S. STAG: species tree inference from all genes. bioRxiv. 2018;267914. https://doi.org/10.1101/267914.

Emms DM, Kelly S. OrthoFinder: phylogenetic orthology inference for comparative genomics. Genome Biol. 2019;20(1):238–238.

Erper I, Agustí-Brisach C, Tunali B, Armengol J. Characterization of root rot disease of kiwifruit in the Black Sea region of Turkey. Eur J Plant Patho. 2013;136(2):291–300.

- Garrison E, Sirén J, Novak AM, Hickey G, Eizenga JM, Dawson ET, et al. Variation graph toolkit improves read mapping by representing genetic variation in the reference. Nat Biotechnol. 2018;36(9):875–9.
- Guo C, Luo Y, Gao LM, Yi TS, Li HT, Yang JB, et al. Phylogenomics and the flowering plant tree of life. J Integr Plant Biol. 2023;65(2):299–323.

Han X, Zhang Y, Zhang Q, Ma N, Liu X, Tao W, et al. Two haplotype-resolved, gap-free genome assemblies for *Actinidia latifolia* and *Actinidia chinensis* shed light on the regulatory mechanisms of vitamin C and sucrose metabolism in kiwifruit. Mol Plant. 2023;16(2):452–70.

Huang H. Kiwifruit. San Diego: Academic Press; 2016.

Huang H, Li J, Lang P, Wang S. Systematic relationships in actinidia as revealed by cluster analysis of digitized morphological descriptors. Acta Hortic. 1999;498:71–8.

Huang H, Gong J, Wang S, He Z, Zhang Z, Li JQ. Genetic diversity in the genus *Actinidia*. Chinese Biodiversity. 2000;8:1–12.

Huang S, Ding J, Deng D, Tang W, Sun H, Liu D, et al. Draft genome of the kiwifruit *Actinidia chinensis*. Nat Commun. 2013;4:2640.

Isner JC, Xu Z, Costa JM, Monnet F, Batstone T, Ou X, et al. Actin filament reorganisation controlled by the SCAR/WAVE complex mediates stomatal response to darkness. New Phytol. 2017;215(3):1059–67.

Kabir N, Wang X, Lu L, Qanmber G, Liu L, Si A, et al. Functional characterization of *TBL* genes revealed the role of *GhTBL7* and *GhTBL58* in cotton fiber elongation. Int J Biol Macromol. 2023;241: 124571.

Kim D, Paggi JM, Park C, Bennett C, Salzberg SL. Graph-based genome alignment and genotyping with HISAT2 and HISAT-genotype. Nat Biotechnol. 2019;37(8):907–15.

Kovaka S, Zimin AV, Pertea GM, Razaghi R, Salzberg SL, Pertea M. Transcriptome assembly from long-read RNA-seq alignments with StringTie2. Genome Biol. 2019;20(1):278–278.

Kubátová Z, Pejchar P, Potocký M, Sekereš J, Žárský V, Kulich I. Arabidopsis trichome contains two plasma membrane domains with different lipid compositions which attract distinct EXO70 subunits. Int J Mol Sci. 2019;20(15):3803.

Kulich I, Vojtíková Z, Glanc M, Ortmannová J, Rasmann S, Žárský V. Cell wall maturation of Arabidopsis trichomes is dependent on exocyst subunit EXO70H4 and involves callose deposition. Plant Physiol. 2015;168(1):120–31.

Lee MM, Schiefelbein J. WEREWOLF, a MYB-related protein in *Arabidopsis*, is a position-dependent regulator of epidermal cell patterning. Cell. 1999;99(5):473–83.

Lefort V, Desper R, Gascuel O. FastME 2.0: a comprehensive, accurate, and fast distance-based phylogeny inference program. Mol Biol Evol. 2015;32:2798–800.

Li JQ, Cai Q, Huang HW. On the phylogeny of the genus *Actinidia* Lindley. In: Huang HW, editor. Advances in *Actinidia* research. Beijing: Science Press; 2000. p. 80–6.

Li P, Quan X, Jia G, Xiao J, Cloutier S, You FM. RGAugury: a pipeline for genomewide prediction of resistance gene analogs (RGAs) in plants. BMC Genomics. 2016;17(1):852.

Li X, Zeng S, Liu J, Wang Y, Sui Y. Introduction and multiplex management strategies of postharvest fungal diseases of kiwifruit: A review. Biol Control. 2022;176:105096.

Li CC, Bao Y, Hou T, Li JC, Ma ZY, Wang N, et al. Insights into chloroplast genome evolution in Rutaceae through population genomics. Hortic Adv. 2024a;2:13.

Li X, Huo L, Li X, Zhang C, Gu M, Fan J, et al. Genomes of diverse *Actinidia* species provide insights into *cis*-regulatory motifs and genes associated with critical traits. BMC Biol. 2024b;22(1):200. Liao G, Chen L, He Y, Li X, Lv Z, Yi S, et al. Three metabolic pathways are responsible for the accumulation and maintenance of high AsA content in kiwifruit (*Actinidia eriantha*). BMC Genomics. 2021;22(1):13.

Liu Y, Li D, Zhang Q, Song C, Zhong C, Zhang X, et al. Rapid radiations of both kiwifruit hybrid lineages and their parents shed light on a two-layer mode of species diversification. New Phytol. 2017;215(2):877–90.

Liu Y, Du H, Li P, Shen Y, Peng H, Liu S, et al. Pan-genome of wild and cultivated soybeans. Cell. 2020;182(1):162–76.

Liu X, Xie X, Zhong C, Li D. Comparative transcriptome analysis revealed the key genes regulating ascorbic acid synthesis in *Actinidia*. Int J Mol Sci. 2021;22(23):12894.

Liu X, Wu R, Bulley SM, Zhong C, Li D. Kiwifruit MYBS1-like and GBF3 transcription factors influence l-ascorbic acid biosynthesis by activating transcription of GDP-L-galactose phosphorylase 3. New Phytol. 2022;234(5):1782–800.

Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. Genome Biol. 2014;15:550.

Lu X, Yu X, Li G, Qu M, Wang H, Liu C, et al. Genome assembly of autotetraploid *Actinidia arguta* highlights adaptive evolution and dissects important economic traits. Plant Commun. 2024;5(6):100856.

Lyu X, Xia Y, Wang C, Zhang K, Deng G, Shen Q, et al. Pan-genome analysis sheds light on structural variation-based dissection of agronomic traits in melon crops. Plant Physiol. 2023;193(2):1330–48.

Majoros WH, Pertea M, Salzberg SL. TigrScan and GlimmerHMM: two open source *ab initio* eukaryotic gene-finders. Bioinformatics. 2004;20(16):2878–9.

Manni M, Berkeley MR, Seppey M, Simão FA, Zdobnov EM. BUSCO update: Novel and streamlined workflows along with broader and deeper phylogenetic coverage for scoring of eukaryotic, prokaryotic, and viral genomes. Mole Biol Evol. 2021;38(10):4647–54.

Marcais G, Delcher A, Phillippy A, Coston R, Salzberg S, Zimin A. MUMmer4: A fast and versatile genome alignment system. PLoS Comput Biol. 2018;14:e1005944.

Miao T, Bao H, Ling H, Li P, Zhang Y, He Y, et al. Comparative transcriptomic analysis revealed the suppression and alternative splicing of kiwifruit (*Actinidia latifolia*) NAP1 Gene mediating trichome development. Int J Mol Sci. 2023;24(5):4481.

Michelotti V, Lamontanara A, Buriani G, Orrù L, Cellini A, Donati I, et al. Comparative transcriptome analysis of the interaction between *Actinidia chinensis* var. *chinensis* and *Pseudomonas syringae* pv. *actinidiae* in absence and presence of acibenzolar-S-methyl. BMC Genomics. 2018;19(1):585.

Minh BQ, Schmidt HA, Chernomor O, Schrempf D, Woodhams MD, von Haeseler A, et al. IQ-TREE 2: New models and efficient methods for phylogenetic inference in the genomic era. Mol Biol Evol. 2020;37(5):1530–4.

Pereira C, Costa P, Pinheiro L, Balcão VM, Almeida A. Kiwifruit bacterial canker: an integrative view focused on biocontrol strategies. Planta. 2021;253(2):49.

Pilkington SM, Crowhurst R, Hilario E, Nardozza S, Fraser L, Peng Y, et al. A manually annotated *Actinidia chinensis* var. *chinensis* (kiwifruit) genome highlights the challenges associated with draft genomes and gene prediction in plants. BMC Genomics. 2018;19(1):257.

Purcell S, Neale B, Todd-Brown K, Thomas L, Ferreira MAR, Bender D, et al. PLINK: A tool set for whole-genome association and population-based linkage analyses. Am J Hum Genet. 2007;81(3):559–75.

Qin P, Lu H, Du H, Wang H, Chen W, Chen Z, et al. Pan-genome analysis of 33 genetically diverse rice accessions reveals hidden genomic variations. Cell. 2021;184(13):3542–58.

Shang L, Li X, He H, Yuan Q, Song Y, Wei Z, et al. A super pan-genomic landscape of rice. Cell Res. 2022;32(10):878–96.

Song Y, Sun L, Lin M, Chen J, Qi X, Hu C, et al. Comparative transcriptome analysis of resistant and susceptible kiwifruits in response to *Pseudomonas syringae* pv. *Actinidiae* during early infection Plos ONE. 2019;14(2):e0211913.

Stamatakis A. RAxML version 8: a tool for phylogenetic analysis and postanalysis of large phylogenies. Bioinformatics. 2014;30(9):1312–3.

Stanke M, Diekhans M, Baertsch R, Haussler D. Using native and syntenically mapped cDNA alignments to improve de novo gene finding. Bioinformatics. 2008;24(5):637–44. Sutherland PW, Fullerton CG, Schröder R, Hallett IC. Cell wall changes in *Actinidia arguta* during softening. Sci Hortic. 2017;226:173–83.

- Tahir J, Hoyte S, Bassett H, Brendolise C, Chatterjee A, Templeton K, et al. Multiple quantitative trait loci contribute to resistance to bacterial canker incited by *Pseudomonas syringae* pv. *actinidiae* in kiwifruit (*Actinidia chinensis*). Hortic Res. 2019;6:101.
- Tahir J, Crowhurst R, Deroles S, Hilario E, Deng C, Schaffer R, et al. First chromosome-scale assembly and deep floral-bud transcriptome of a male kiwifruit. Front Genet. 2022;13: 852161.
- Tang H, Bowers JE, Wang X, Ming R, Alam M, Paterson AH. Synteny and collinearity in plant genomes. Science. 2008;320(5875):486–8.
- Tang W, Sun X, Yue J, Tang X, Jiao C, Yang Y, et al. Chromosome-scale genome assembly of kiwifruit *Actinidia eriantha* with single-molecule sequencing and chromatin interaction mapping. GigaScience. 2019;8(4):giz27.
- Tirnaz S, Bayer PE, Inturrisi F, Zhang F, Yang H, Dolatabadian A, et al. Resistance gene analogs in the *Brassicaceae*: identification, characterization, distribution, and evolution. Plant Physiol. 2020;184(2):909–22.
- Van de Weyer AL, Monteiro F, Furzer OJ, Nishimura MT, Cevik V, Witek K, et al. A species-wide inventory of NLR genes and alleles in *Arabidopsis thaliana*. Cell. 2019;178(5):1260–72.
- Van Wersch S, Li X. Stronger when together: Clustering of plant NLR disease resistance genes. Trends Plant Sci. 2019;24(8):688–99.
- Wang S, Hubbard L, Chang Y, Guo J, Schiefelbein J, Chen JG. Comprehensive analysis of single-repeat R3 MYB proteins in epidermal cell patterning and their transcriptional regulation in Arabidopsis. BMC Plant Biol. 2008;8:81.
- Wang K, Li M, Hakonarson H. ANNOVAR: functional annotation of genetic variants from high-throughput sequencing data. Nucleic Acids Res. 2010;38(16): e164.
- Wang F, Li J, Ye K, Liu P, Gong H, Jiang Q, et al. An in vitro Actinidia bioassay to evaluate the resistance to *Pseudomonas syringae* pv. actinidiae. Plant Pathol J. 2019;35:372–80.
- Wang T, Jia ZH, Zhang JY, Liu M, Guo ZR, Wang G. Identification and analysis of NBS-LRR genes in *Actinidia chinensis* genome. Plants (Basel). 2020b;9(10):1350.
- Wang F, Mo Q, Ye K, Gong H, Qi B, Liu P, et al. Evaluation of the wild Actinidia germplasm for resistance to Pseudomonas syringae pv. actinidiae. Plant Pathol. 2020a;69(6):979–89.
- Wang B, Luo Q, Li Y, Du K, Wu Z, Li T, et al. Structural insights into partner selection for MYB and bHLH transcription factor complexes. Nat Plants. 2022;8(9):1108–17.
- Wang T, Duan S, Xu C, Wang Y, Zhang X, Xu X, et al. Pan-genome analysis of 13 Malus accessions reveals structural and sequence variations associated with fruit traits. Nat Commun. 2023a;14(1):7377.
- Wang Y, Dong M, Wu Y, Zhang F, Ren W, Lin Y, et al. Telomere-to-telomere and haplotype-resolved genome of the kiwifruit *Actinidia eriantha*. Mol Hortic. 2023b;3(1):4.
- Wang Y, Li P, Zhu Y, Zhang F, Zhang S, He Y, et al. Graph-based pangenome of Actinidia chinensis reveals structural variations mediating fruit degreening. Adv Sci. 2024;11(28):e2400322.
- Wu H, Ma T, Kang M, Ai F, Zhang J, Dong G, et al. A high-quality *Actinidia chinensis* (kiwifruit) genome. Hortic Res. 2019;6(1):117.
- Wu Y, Li D, Hu Y, Li H, Ramstein GP, Zhou S, et al. Phylogenomic discovery of deleterious mutations facilitates hybrid potato breeding. Cell. 2023;186(11):2313–28.
- Xia H, Deng H, Li M, Xie Y, Lin L, Zhang H, et al. Chromosome-scale genome assembly of a natural diploid kiwifruit (*Actinidia chinensis* var. *deliciosa*). Sci Data. 2023;10(1):92.
- Yan H, Sun M, Zhang Z, Jin Y, Zhang A, Lin C, et al. Pangenomic analysis identifies structural variation associated with heat tolerance in pearl millet. Nat Genet. 2023;55(3):507–18.
- Yang Z. PAML 4: phylogenetic analysis by maximum likelihood. Mol Biol Evol. 2007;24(8):1586–91.
- Yang Y, Chen L, Wang C, Peng H, Yin W, Li R, et al. Pathogenic fungi diversity of "CuiXiang" kiwifruit black spot disease during storage. Horticulturae. 2022;8(1):13.
- Yao X, Wang S, Wang Z, Li D, Jiang Q, Zhang Q, et al. The genome sequencing and comparative analysis of a wild kiwifruit *Actinidia eriantha*. Mol Hortic. 2022;2(1):13.

- Yu X, Qin M, Qu M, Jiang Q, Guo S, Chen Z, et al. Genomic analyses reveal dead-end hybridization between two deeply divergent kiwifruit species rather than homoploid hybrid speciation. Plant J. 2023;115(6):1528–43.
- Yue J, Chen Q, Wang Y, Zhang L, Ye C, Wang X, et al. Telomere-to-telomere and gap-free reference genome assembly of the kiwifruit *Actinidia chinensis*. Hortic Res. 2023;10(2):uhac264.
- Yue J, Chen Q, Zhang S, Lin Y, Ren W, Li B, et al. Origin and evolution of the kiwifruit Y chromosome. Plant Biotechnol J. 2024;22(2):287–9.
- Zhang X, Guo R, Shen R, Landis JB, Jiang Q, Liu F, et al. The genomic and epigenetic footprint of local adaptation to variable climates in kiwifruit. Hort Res. 2023;10:uhad031.
- Zhang F, Wang Y, Lin Y, Wang H, Wu Y, Ren W, et al. Haplotype-resolved genome assembly provides insights into evolutionary history of the *Actinidia arguta* tetraploid. Mol Hortic. 2024;4(1):4.
- Zhao Q, Shi Y, Xu C, Jiang Z, Liu J, Sui Y, et al. Control of postharvest blue and gray mold in kiwifruit by *Wickerhamomyces anomalus* and its mechanism of antifungal activity. Postharvest Biol Tec. 2023;201: 112345.
- Zheng Z, Zhu M, Zhang J, Liu X, Hou L, Liu W, et al. sequence-aware merger of genomic structural variations at population scale. Nat Commun. 2024;15(1):960.

Publisher's Note

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.