ORIGINAL ARTICLE



Structural variations and environmental specificities of flowering time-related genes in *Brassica napus*

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Received: 13 September 2022 / Accepted: 16 February 2023 / Published online: 10 March 2023 © The Author(s), under exclusive licence to Springer-Verlag GmbH Germany, part of Springer Nature 2023

Abstract

Key message We found that the flowering time order of accessions in a genetic population considerably varied across environments, and homolog copies of essential flowering time genes played different roles in different locations. Abstract Flowering time plays a critical role in determining the life cycle length, yield, and quality of a crop. However, the allelic polymorphism of flowering time-related genes (FTRGs) in *Brassica napus*, an important oil crop, remains unclear. Here, we provide high-resolution graphics of FTRGs in *B. napus* on a pangenome-wide scale based on single nucleotide polymorphism (SNP) and structural variation (SV) analyses. A total of 1337 FTRGs in *B. napus* were identified by aligning their coding sequences with *Arabidopsis* orthologs. Overall, 46.07% of FTRGs were core genes and 53.93% were variable genes. Moreover, 1.94%, 0.74%, and 4.49% FTRGs had significant presence-frequency differences (PFDs) between the spring and semi-winter, spring and winter, and winter and semi-winter ecotypes, respectively. SNPs and SVs across 1626 accessions of 39 FTRGs underlying numerous published qualitative trait loci were analyzed. Additionally, to identify FTRGs specific to an eco-condition, genome-wide association studies (GWASs) based on SNP, presence/absence variation (PAV), and SV were performed after growing and observing the flowering time order (FTO) of plants in a collection of 292 accessions at three locations in two successive years. It was discovered that the FTO of plants in a genetic population changed a lot across various environments, and homolog copies of some key FTRGs played different roles in different locations. This study revealed the molecular basis of the genotype-by-environment (G×E) effect on flowering and recommended a pool of

Communicated by Isobel AP Parkin.

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candidate genes specific to locations for breeding selection.

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Introduction

Rapeseed (Brassica napus L.) (AACC, n = 19), which originated from interspecific hybridization between Brassica rapa (AA, n = 10) and Brassica oleracea (CC, n = 9) approximately 7500 years ago, is a major source of edible oil and livestock feed rich in protein (Chalhoub et al. 2014; Nagaharu 1935). Allotetraploid rapeseed has been used in some major production regions, such as China and Canada, for less than a century. It has widely adapted to adverse climate zones across Eurasia, North America, and Australia by developing a broad variation in flowering time (FT) from early spring to late winter types. Therefore, variations in FT of rapeseed are crucial for its adaptation to new environments. Moreover, FT plays a critical role in determining the length of the life cycle period, yield, and quality of seeds and is one of the major targets for rapeseed breeding programs (Guo et al. 2014).

Flowering is regulated by a variety of endogenous and environmental cues such as changes in hormone levels (e.g., gibberellin, GA), age, carbohydrate status, ambient temperature, and photoperiod. Flowering is promoted by florigen, a mobile signal substance encoded by the floral integrator gene Flowering LOCUS T (FT). Florigens activate floral identity genes in the shoot apical meristem (SAM) and induce floral organ development (Freytes et al. 2021; Kinoshita and Richter 2020). In Arabidopsis, the CONSTANS (CO)-FT regulon genes are involved in various pathways that regulate flowering. Over the past decades, key regulators of the main flowering pathways have been identified, such as microRNA156 which targets SQUAMOSA PROMOTER BINDING-LIKE (SPL), a flowering promoter, in age-related pathways (Hyun et al. 2017); FLOWERING CONTROL LOCUS A (FCA) and LUMINIDEPENDENS (LD) in the autonomous pathways (Chakrabortee et al. 2016; Macknight et al. 1997); GIBBER-ELLIC ACID INSENSITIVE (GAI), REPRESSOR OF GA (RGA), and SPINDLY (SPY) in the gibberellic acid (GA) pathway (Jacobsen and Olszewski 1993; Peng and Harberd 1993; Tyler et al. 2004); CO, FT, DE-ETIOLATED 1 (DET1), and PHYTOCHROME A (PHYA) in the photoperiod pathway (Li et al. 2015; Putterill et al. 1995; Whitelam et al. 1993; Yoo et al. 2005); and FLOWERING LOCUS C (FLC), VERNALIZATION 1 (VRN1), and VRN2 in the vernalization pathway (Gendall et al. 2001; King et al. 2013; Sharma et al. 2020). A large number of these regulators are transcriptional factors (TFs), cofactors for TFs that have numerous downstream targeting genes, and chromatin remodelers. Additionally, hundreds of genes influence FT in Arabidopsis. Due to its much larger genome, B. napus, a relative of Arabidopsis in the family Brassicaceae, possesses 3-6 times more flowering time-related genes (FTRGs) in its allotetraploid genome than Arabidopsis.

Owing to advances in sequencing technology such as genomics, genome assembly and annotation have considerably improved (Bayer et al. 2021). The first reference genome of *B. napus*, the Darmor-*bzh* assembly, was published in 2014 (Chalhoub et al. 2014). The genomes of two winter ecotype accessions, Tapidor and Express 617 (Bayer et al. 2017; Lee et al. 2020), and those of two semi-winter ecotype accessions, ZS11 and Ningyou-7 (Chen et al. 2021; Zou et al. 2019), were de novo assembled and annotated in succession. However, none of these single reference genomes can represent the entire gene content of B. napus, since structural variations (SVs), such as gene presence and absence variations (PAVs) and copy number variations (CNVs), exist among these plants. To overcome this limitation, a pangenome for rapeseed was constructed and analyzed for SVs that were associated with agronomic traits (Song et al. 2020), disease resistance (Dolatabadian et al. 2020), and gene-loss propensity by interspecific hybridization (Bayer et al. 2021).

 Table 1
 The total number of accessions and proportion of ecotypes in each experiment

	PAV analysis	SV analysis	GWAS multiple environ
Total	1626***	991***	292***
Spring	247 (15.19%)	188 (18.97%)	58 (19.86%)
Winter	712 (43.79%)	658 (66.40%)	180 (61.64%)
Semi-winter	667 (41.02%)	145 (14.63)	54 (18.49%)

***Please refer to Table S23, Table S1 in Wu et al. (2019), and Table S8 for the ID of the1626, 991, and 292 accessions, respectively

In this study, we aimed to improve the resolution of the FTRG panorama of B. napus on a pangenome-wide scale, detect single nucleotide polymorphisms (SNPs) associated with FTRGs specific to ecotypes, and identify SVs across a collection of germplasms belonging to different ecotypes. Finally, we performed multiple-year and multiple-location experiments to determine FTRGs associated with flowering time order variations (FTOV) specific to geographic locations and annual climate conditions by observing the changes in flowering time rank (FTR) of plants and performing genome-wide association studies (GWAS) based on SNPs, PAVs, and SVs in a genetic population consisting of 292 accessions. The results of this study provide insights into the FTRG network of B. napus and the molecular nature of a genotype-by-environment $(G \times E)$ effect and provide a reference to select and/or manipulate candidate FTRGs specific to geographic locations for introduction and domestication.

Materials and methods

Number of accessions investigated in each analysis

We investigated 1626, 991, and 292 accessions in PAV analysis, SV analysis, and multiple-environment GWAS, respectively (Table 1). The ID of these accessions and the proportion of ecotypes in each analysis are indicated or noted in the table.

Pangenome

The pangenome of *B. napus* constructed by Song et al. (2021) was used in this study. ZS11 was used as the reference genome, and 1688 resequenced genomes were combined to construct the pangenome (Lu et al. 2019; Parkin et al. 2005; Song et al. 2020; Wang et al. 2018; Wu et al. 2019). A total of 62 accessions were removed due to the deficiency of ecotype information. The ID of the accessions and the ecotypes to which they belong are provided in Table S23.

Identification of candidate FTRGs

Sequences of *Arabidopsis* FTGRs were downloaded from the flowering interaction database (http://www.phytosyste ms.ulg.ac.be/florid) (Bouché et al. 2016). The coding sequences (CDSs) of the *Brassica napus* pangenome were downloaded from the BnPIR database (http://cbi.hzau.edu. cn/bnapus/) (Song et al. 2021). The CDSs were compared with *Arabidopsis* FTRGs using blastn with an e-value of < 1e-20 and identity of > 80%.

SNP discovery

The SNP information for *B. napus* was downloaded from BnPIR, and the variations were annotated using VEP v99 (McLaren et al. 2016).

SV identification and selection analysis

Fastp (Chen et al. 2018) was used to filter out low-quality sequences from the raw data created by Wu et al. (2019) The ID of the 991 accessions and the ecotypes to which they belong are listed in Table S23. Clean data were mapped to the ZS11 reference genome using bowtie2 (Langmead and Salzberg 2012). BreakDancer and Delly were used to identify SVs (Chen et al. 2009). The results of BreakDancer were converted into VCF format. Then, SURVIVOR (version 1.0.6) (Jeffares et al. 2017) was used to identify SVs detected by both tools using the following parameters: "SURVIVOR merge Name 1000 2 1 1 0 50." SURVIVOR defines and merges SVs based on breakpoints, SV types, and distances between SV chains. The SVs of all samples were merged using bcftools (Danecek et al. 2021). When merging, the depth of the reads at the breakpoints of SVs must be $\geq 4x$, the distance between the start and end points of two adjacent SVs should not exceed 1000 bases, and the length of the overlap between the two SVs should > 50% of the total length.

To analyze SV under selection, the occurrence frequencies of each SV were calculated in the spring, winter, and semi-winter groups. Similar to the gene PAV analysis, the significance of the differences in the frequencies for each SV between groups was determined using Fisher's exact test. *p*-values were then corrected using the Benjamini and Hochberg method. SVs under selection were identified with a false discovery rate (FDR) < 0.001 and a fold-change > 2. We set a 2-kb interval upstream of the genes as the promoter regions. Based on genome annotation and SV locations, we evaluated the effect of putative SV on the gene, CDS, and promoter regions. The SVs under selection were then analyzed to identify the associated genes.

Gene PAV selection analysis

The PAV of the pangenome was downloaded from the BnPIR database (http://cbi.hzau.edu.cn/bnapus/) (Song et al. 2021). Then, the frequencies of each gene in different *B. napus* ecotypes (spring, winter, and semi-winter) were calculated. The significance of the differences in gene frequencies between the groups was determined using Fisher's exact test. *p*-values were corrected using the Benjamini and Hochberg method to obtain the adjusted *p*-value. Genes with a fold-change in frequency > 2 between the two groups and a false discovery rate value <0.001 were defined as genes under gene PAV selection.

The protein sequences of all genes in the *B. napus* pangenome were compared with those in the UniPort and Kyoto Encyclopedia of Genes and Genomes (KEGG) databases using BLASTP with a threshold of 1E-5. Using the Retrieve/ID mapping tool (https://www.uniprot.org/uploa dlists/), UniPort IDs were converted to GO IDs to perform Gene Ontology (GO) annotation for all genes in the *B. napus* pangenome. Based on the GO and KEGG annotations, a hypergeometric test was used to perform functional enrichment analysis of the genes under gene PAV selection. The dhyper function in R was used to calculate *p*-values in the enrichment analysis. Additionally, the multiple-test corrected *p*-value (*q*-value) was calculated using the Benjamini–Hochberg function implemented in the R. *q*-value 0.05 was considered statically significant.

Linking known QTL and FTRGs

Chromosomal coordinates of QTL intervals for FT were obtained from previous studies (Jian et al. 2019; Liu et al. 2022; Scheben et al. 2020; Xu et al. 2021). The coordinates of the QTL intervals were converted to coordinates of the ZS11 genome.

Plots and graphs

Waterfall plots were drawn using GenVisR v1.11.3 (Skidmore et al. 2016). PAV information of FTRGs was used to generate a heatmap using R (version 4.0.3) with the ComplexHeatmap package (version 2.6.2, https://bioconductor. org/packages/release/bioc/html/ComplexHeatmap.html) (Gu et al. 2016). Scatter and line plots were drawn using the R package ggplot2 (Villanueva and Chen 2019). Sankey plots were drawn using the R package networkD3 (Allaire et al. 2017).

Construction of core germplasm

The 292 accessions for GWAS were selected based on the phylogenetic tree and principal component analyses of the

genetic diversity of the 991 accessions. This GWAS population represented around 97.0% SNPs and 97.0% InDel polymorphisms of the large population consisting of 991 global accessions, which were sequenced on a genome-wide scale in our previous study (Wu et al. 2019).

Plant growth conditions

The 292 core accessions were grown at three locations, namely HZ (30.89° N and 119.63° E), JX (30.86° N and 120.70° E), and XY (34.80° N and 108.10° E), in two successive years during rapeseed growing seasons in 2019–2021. Sixteen plants from each accession were planted in a block (120×100 cm) with three replicates. Field management measures were carried out in accordance with the local rapeseed planting practices.

Observation of FT

The FT date for each accession was recorded when 50% of the plants had visible open flowers on their main inflorescence. The DSF was then calculated based on the number of days between sowing dates and FT dates.

GWAS-SNP, GWAS-PAV, and GWAS-SV

High-quality SNPs with minor allele frequency (MAF) > 0.05 discovered in the core accessions were used for GWAS-SNP. GWAS-SNP was performed using the BnaGVD website (http://rapeseed.biocloud.net/home) (Yan et al. 2021) with the default parameters of the EMMAX model. Associated genes were searched for in 75-kb sequence regions adjacent to SNPs. The SV and gene PAV information was used as the genotype instead of the SNP to perform GWAS-SV and GWAS-PAV using FarmCPU (default parameter) (Liu et al. 2016) in rMVP, with the threshold of significance set at 0.05/SV or 0.05/PAV number. SNPs were used for PCA using GCTA (Yang et al. 2011), and SNP-based principal components were used as covariates.

Results

PAV of B. napus FTRGs

In this study, we used the orthologous genes of *Arabidopsis thaliana* as a reference and identified 1337 FTRGs in the *B. napus* pangenome. Of these, 616 (46.07%) were core genes and 721 (53.93%) were variable genes; 482 (36.05%) genes (207 core and 275 variable genes) and 537 (40.16%) genes (310 core and 227 variable genes) were distributed in the A and C subgenomes, respectively. A total of 297 (22.21%)

genes (all variables) were assigned to the contigs of the pangenome in addition to the two subgenomes. A group of 21 (1.57%) (two core and 19 variable) genes were identified in unplaced contigs (Table 2).

The distribution of the FTRGs in the reference genome is shown in Fig. 1. The largest number of FTRGs (99 genes) was identified on chromosome C03, whereas the smallest number of FTRGs (33 genes) was identified on chromosomes A04, A08, and A10. For each chromosome in the A genome, more variable FTRGs were identified than core FTRGs. In the C genome, except for C02 and C09, fewer variable FTRGs were identified than core FTRGs on most chromosomes. Chromosome C03 had the highest frequency (86.79%) of core FTRGs, whereas chromosome C02 had the lowest frequency (30.19%). Detailed information on variable and core FTRGs in the pangenome is provided in Table S1.

PAV-based pairwise comparisons were performed to analyze the presence-frequency difference (PFD) of FTRGs among the three ecotypes. The comparison of PFDs between the spring and winter ecotypes revealed five significantly more frequent and five significantly less frequent FTRGs in the spring ecotype than in the winter ecotype (Fig. 2a, d). Similarly, three FTRGs were significantly more frequent and 23 FTRGs were significantly less frequent in the spring ecotype than in the semi-winter ecotype (Fig. 2b, e). Moreover, six FTRGs were significantly more frequent in the winter ecotype than in the semi-winter ecotype and 54 FTRGs were significantly less frequent in the winter ecotype than in the semi-winter ecotype than in the semi-winter ecotype (Fig. 2c, f). The PFDs of the FTRGs in the ecotypes are provided in Table S2.

Collectively, 26 (1.94%), 10 (0.74%), and 60 (4.49%) FTRGs had significant PFDs between the spring and semiwinter, spring and winter, and winter and semi-winter ecotypes, respectively. Excluding duplicates, 75 (5.61%) FTRGs had PFDs among the three ecotypes. The IDs of the PFD genes are listed in Table S3, and the overlapping numbers of the FTRGs between the ecotypes are indicated in a Venn diagram (Fig. S1). To intuitively overview the PAV status of each FTRG across the 1626 individuals (Table S4), we drew a fine-resolution $(75 \times 1626 \text{ pixels})$ heatmap (Fig. 3). As shown in Fig. 3, most FTRGs present in the semi-winter ecotype were absent in the spring and winter ecotypes, and seven FTRGs, namely BnaSORG0131600UN (PROTEIN ARGININE METHYLTRANSFERASE 4A, PRMT4A), BnaA02G0022900ZS (UBIQUITIN-SPECIFIC PROTEASE 12, UBP12), BnaA06G0299300ZS (CYCLIN-DEPENDENT KINASE C2, CDKC2), BnaA10G0150400ZS (SENSITIVITY TO RED LIGHT REDUCED 1, SRR1), BnaA03G0188600ZS (NUCLEAR FACTOR Y SUBUNIT B1, NF-YB1), BnaA03G0194600ZS (EARLY IN SHORT DAYS 6, ESD6), and BnaA07G0171100ZS (CYCLING DOF FACTOR 2, CDF2), were mostly different across the three ecotypes in terms of PAV.

Table 2The number anddistribution of rapeseedflowering time-related genes(FTRGs) in the pangenome

Genome region	Chromosome	Variable FTRG number	Core FTRG number	All FTRG number
Reference A genome	A01	23	18	41
	A02	34	19	53
	A03	40	32	72
	A04	18	15	33
	A05	26	23	49
	A06	36	26	62
	A07	22	16	38
	A08	18	15	33
	A09	36	32	68
	A10	22	11	33
	Total A genome	275	207	482
Reference C genome	C01	22	26	48
	C02	37	16	53
	C03	38	61	99
	C04	28	34	62
	C05	7	46	53
	C06	19	22	41
	C07	20	42	62
	C08	17	30	47
	C09	39	33	72
	Total C genome	227	310	537
Reference A and C genome		502	517	1019
Pangenome additional contigs		297	0	297
Reference genome unplaced contigs		19	2	21
Pangenome		721	616	1337

SV analyses of FTRGs

We previously resequenced a global collection of 991 accessions and investigated the SNPs for genomic polymorphisms (Table S23; Wu et al. 2019). Here, we identified 124,287 SVs, including insertions/deletions (indels), inversions, and duplications in the collection for the subsequent GWAS analyses (Table 3). Most SVs were of low frequency (<0.01) (Fig. S2). Of these, 55,662 SVs were identified in the A genome, 68,565 SVs were identified in the C genome, and 60 SVs were identified in unplaced contigs in the ZS11 reference genome. On average, more SVs per chromosome were identified in the C genome (7618.3) than in the A genome (5566.2). Most SVs were indels (96.95% overall), varying from 95.53% to 98.31% on each chromosome (Table 3).

To identify the SVs that contributed to the differentiation of the three ecotypes, we compared the frequency of each SV among the spring, winter, and semi-winter ecotypes. A total of 1867 SVs with significantly different frequencies between ecotypes were identified. The position, SV type, length, allelic frequency, and pairwise ecotype comparisons are shown in detail in Table S5. The comparison results revealed five SVs that had significantly higher frequencies and 31 SVs that had significantly lower frequencies in the spring ecotype than in the semi-winter ecotype (Figs. S3 and S4). Moreover, 524 SVs with significantly higher frequencies and 263 SVs with significantly lower frequencies in the spring ecotype than in the winter ecotype were identified (Figs. S5 and S6). The results of the comparison of the SVs between the winter and semiwinter ecotypes revealed 187 SVs with significantly higher frequencies and 971 SVs with significantly lower frequencies in the winter ecotype than in the semi-winter ecotype (Figs. S7 and S8). Notably, 52 SVs had significantly different frequencies among the three ecotypes, which might have contributed to the differentiation of the ecotypes. The position, SV type, length, allelic frequency, pairwise ecotype comparisons, and IDs of the SV-associated FTRGs are summarized in Table S6. Collectively, we identified 3204 SVs on FTRGs in B. napus. Of these, 61.77% were indels, 38.14% were inversions, and only 0.06% were duplications.

Analysis of known quantitative trait loci (QTL) that regulates FT

Numerous QTL that control FT were mapped in *B. napus* (Jian et al. 2019; Liu et al. 2022; Scheben et al. 2020; Xu



Fig. 1 The distribution of dispensable flowering time-related genes (FTRGs) and all FTRGs across the reference genomes. The *Y*-axis represents gene densities, which were normalized to the genome-wide

maximum of each measurement peaking at 1. The *X*-axis stands for the ID and size of each chromosome of the genome

et al. 2021). These QTL contributed considerably to flowering time variations. To investigate the FTRGs underlying the known QTL, we remapped the QTL to the pangenome. SNPs, PAVs, and SVs of the FTRGs underlying 14 QTL were analyzed. Seventy-eight FTRGs underlying those QTL were identified. Of these, 39 FTRGs varied significantly across ecotypes. A waterfall plot was drawn to display the variant types, such as gene-loss, stop-codon-loss, missense, synonymous, intron, splice-region, 3'-UTR variants, and 5'-UTR variants of the FTRGs underlying the QTL (Fig. 4). Among these FTRGs, *BnaA06G0332400ZS (FLOWER-ING-PROMOTING FACTOR 1, FPF1)* had the lowest mutation rate, and the mutations were mainly synonymous. *BnaA06G0277300ZS (CYCLIC DOF FACTOR 1-LIKE, CDF1)* had the highest mutation rate and the mutations were gene-loss and missense variants, indicating considerably high genetic polymorphism and functional diversity. Although *BnaA06G0126600ZS* (*ACTIN-RELATED PRO-TEIN 4, ARP4*) had a relatively low mutation rate, the mutations were mostly gene-loss variants, indicating PAVs across the 1626 individuals. *BnaA06G0256500ZS* (*DICER-LIKE 3, DCL3*) was featured with mutations in the stop-codonloss variants. *BnaA06G0270200ZS* (*UBIQUITIN-SPECIFIC PROTEASE 26, UBP26*) had a considerable proportion of synonymous mutations that did not lead to any functional differences. On average, individuals of each ecotype had a similar mutation frequency. However, it was obvious that the differences in mutation frequency between the individuals in the semi-winter type were bigger than those in the winter and spring types. The allelic variation of the 39 FTRGs



Fig. 2 Comparison of the frequencies of flowering time-related genes (FTRGs) between **a** winter and spring, **b** semi-winter and spring, and **c** winter and semi-winter ecotypes. The red and green dots indicate significant differences (p < 0.01) between two comparing pairs, and

they are shown as red and green lines in (d-f). The blue dots in (a-c) were not significantly different between the comparing pairs (color figure online)



Fig. 3 Waterfall plot of presence/absence variation PAV of the flowering time-related genes (FTRGs) among three ecotypes. The IDs of the six FTRGs with the most significant differences between the ecotypes are indicated on the left site of the plot. The blue and yellow

colors of each pixel represent the presence or absence of a particular gene in a genotype, respectively. Accession number = 1626, gene number = 75 (color figure online)

Table 3Number, type, anddistribution of structuralvariations (SVs) identified inthe 991 germplasms population

Chromosome	SV number	SV type				SV location		
		Inversion	Duplication	Deletion	NA	Gene	Promoter	Intergenic
A01	5798	198	20	5570	10	1756	1836	2725
A02	6052	216	23	5795	18	1798	1966	2791
A03	7125	279	11	6825	10	2430	2719	2707
A04	4360	146	13	4191	10	1229	1458	2011
A05	4931	149	17	4762	3	1516	1686	2178
A06	5388	214	17	5147	10	1630	1865	2392
A07	5350	175	17	5151	7	1499	1770	2504
A08	4169	113	10	4039	7	1246	1433	1867
A09	8949	297	36	8606	10	2864	3042	3930
A10	3540	123	13	3400	4	1132	1202	1554
A genome total	55,662	1910	177	53,486	89	17,100	18,977	24,659
C01	6346	156	20	6164	6	1707	1734	3362
C02	8179	194	19	7962	4	2157	2054	4478
C03	10,393	212	8	10,172	1	3064	3112	5045
C04	8665	249	24	8389	3	2525	2265	4498
C05	7029	128	18	6880	3	2107	1935	3566
C06	6297	108	15	6173	1	1722	1607	3344
C07	7367	114	8	7243	2	2018	1913	3911
C08	6266	128	11	6126	1	1784	1743	3222
C09	8023	132	11	7880	0	2190	2101	4291
C genome total	68,565	1421	134	66,989	21	19,274	18,464	35,717
Reference genome unplaced contigs	60	1	32	27	0	0	1	59
Total	124,287	3332	343	120,502	110	36,374	37,442	60,435

across 1626 accessions (including the 991 accessions) is provided in detail in Table S7.

FTOV observation in a genetic population

Based on the genome-wide distribution of FTRGs in B. napus, we attempted to identify the FTRGs associated with flowering time order (FTO) in a given environment. We grew the genetic population consisting of 292 core accessions (Table S23), which were selected based on the resequencing data of 991 global germplasm collections at three locations: Xian-Yang (XY, 34.80°N and 108.10°E), Hu-Zhou (HZ, 30.89°N and 119.63°E), and Jia-Xing (JX, 30.86°N and 120.70°E) in two successive years. The 292 core accessions contains most polymorphisms of 991 populations, and its validity in GWAS analysis was confirmed in previous study. We recorded the days from sowing to flowering (DSF) for each individual in the six environments (location × year). DSF varied from 135 to 194 at HZ, 102 to 202 at JX, and 158 to 213 at XY (Table S8). The number of flowering accessions displayed a bimodal distribution, peaking twice in all six environments, which was particularly clear at XY (Fig. S9). We divided the DSF into five grades, G1 to G5,

corresponding to increasing DSF. We determined the FTR of each accession and arranged the FTO of the population in each environment (Table S8). We drew a Sankey diagram to intuitively show the FTOV across the six environments (Fig. 5). The FTRs of 72 (24.66%) accessions were relatively consistent in various environments. However, the FTRs of 193 (66.10%) accessions fluctuated with changes in FTRs between neighboring grades across the environments. Notably, the FTRs of 27 (9.25%) accessions changed drastically between non-adjacent grades across environments (Fig. 5; Table S8). Therefore, the FTO of a population changes with time and location, and the FTR for a specific accession in two different environments may be considerably different.

GWAS on FT specific to a given environment

To identify the FTRGs associated with FT in a specific environment, we performed GWAS-SNP, GWAS-PAV, and GWAS-SV (excluding PAV). A set of Manhattan plots showed the results of the GWAS-SNP in the six environments (Fig. S10). Genes associated with FT specific to the environment were analyzed (Tables S9–S14). Cross-analyses between locations were performed, and the number of



Fig. 4 Waterfall plot of known qualitative trait locus (QTL) that control flowering time in a recombinant-inbred-line population. The IDs of flowering time-related genes (FTRG) according to the ZS11 reference genome are listed on the left side with histograms indicating the rate of mutation of each gene of the 1626 accessions. The graph on

the top represents the mutation frequency of the accessions across the 39 FTRGs. The different colors of the 39×1626 pixels represent the mutation types as indicated on the right side of the plot (color figure online)

overlapped associated genes is shown in a Venn diagram (Fig. 7a). Eighteen genes were significantly associated with FT at the three locations. We identified 22 and 134 significantly associated genes that overlapped between JX and XY, HZ, and JX, respectively. The numbers of genes that were identified only at HZ, JX, and XY were 758, 1853, and 176, respectively (Table S15). Several associated genes belonged to the 1337 FTRGs that were predicted above (Table S16). Interestingly, homologous gene copies play different roles in various environments. For example, both BnaA-03G0144400ZS and BnaC02G0039100ZS are FLC homologous genes; however, they were identified as associated genes at different locations, that is, BnaA03G0144400ZS was identified in HZ, whereas BnaC02G0039100ZS was identified at JX. Moreover, both BnaC03G0129200ZS and BnaA02G0118200ZS were VERNALISATION INSENSI-TIVE 3 (VIN3) homologs; however, BnaC03G0129200ZS was identified as an associated gene at JX, whereas BnaA-02G0118200ZS was identified at XY (Table S16).

GWAS-PAV and GWAS-SV (excluding PAV) were also performed to supplement the results of GWAS-SNP (Fig. 6). PAV-associated genes in all six environments are provided in Table S17. GWAS-PAV and GWAS-SV (excluding PAV) did not reveal associated genes common to the three locations. GWAS-PAV revealed 11, 17, and 7 genes associated with HZ, JX, and XY, respectively (Figs. 6, 7b). GWAS-SV (excluding PAV) revealed 140, 24, and 3 genes associated with HZ, JX, and XY, respectively (Figs. 6, 7c; Table S18). GWAS-PAV revealed one overlapping gene, BnaA10G0193800ZS (TatD-related DNase), between HZ and JX, and one overlapping gene, BnaA02G0374400ZS (Topless-related Protein), between JX and XY (Fig. 7b). In contrast, GWAS-SV (excluding PAV) revealed other overlapping genes BnaC08G0316200ZS (Spermatogenesis-associated Protein 7, associated with SV114034) between HZ and JX (Fig. 7c). Taken together, these results indicated that GWAS-SNP, GWAS-PAV, and GWAS-SV (excluding PAV) enriched the number of FTRG candidates and provided the Fig. 5 Comparison of flowering time rank in the population comprising 292 accessions across different locations and years. Time is divided into five periods from early to late as shown with the erect columns. The darker the line that links an accession, the more drastic the change in the rank of flowering time of the accession. The lavender color represents the collinearity between years at the same location, and the yellow color represents the collinearity across locations. HZ, JX, and XY indicate the location of field experiments (Hu-Zhou, 30.89°N and 119.63°E; Jia-Xing 30.86°N and 120.70°E; and Xian-Yang 34.80°N and 108.10°E, respectively). Y1 and Y2 stand for the repetition of field experiments in different years (color figure online)



molecular nature underlying the genotype-by-environment $(G \times E)$ effect on FT.

Discussion

The transition from vegetative growth to flowering is an important developmental stage of a plant during its life cycle. FT determines not only the life cycle length but also the yield and quality of a crop. A correct time of flowering is essential in multiple-cropping systems, where one or even two additional cropping seasons for rice production closely follow the harvest of winter crops. The late flowering of the preceding crop delays the planting time of the next crop. Moreover, the cultivation of plants with appropriate FT requires knowledge of allelic variations in FTRGs. In this study, we investigated B. napus FTRGs on a pangenomewide scale and identified SNPs, PAVs, and SVs in a collection of worldwide rapeseed germplasm. Here, we defined SVs as indels, inversions, and duplications, but not PAV of genes (Lu et al. 2019; Song et al. 2020, 2021; Wang et al. 2018; Wu et al. 2019).

The concept of the pangenome was first developed in bacteria, describing an organism's complete genetic material, including core and variable genomes (Tettelin et al. 2005). Various strategies have been developed to improve the pangenome. First, methods were applied to align reads from multiple individuals to a reference genome of good quality, assemble unaligned reads into novel contigs, and add novel contigs to the original reference sequence to construct a pangenome. Second, different strategies were applied to de novo assemble genomes of multiple accessions and align the whole genomes to identify variable genomic regions. More recently, a new strategy was proposed to construct a pangenome graph by aligning whole genomes and storing information on variable regions through the graph (Bayer et al. 2020). In the present study, we used the pangenome constructed by Song et al. (2021), which was developed by mapping 1688 resequenced accessions to the ZS11 reference genome (Lu et al. 2019; Wang et al. 2018; Wu et al. 2019). Dolatabadian et al. (2020) adopted a similar strategy to build a B. napus pangenome by mapping 50 resequenced accessions to the Darmor-bzh reference genome (v8.1). They identified 1749 resistant gene analogs, of which 996 were core, 753 were variable, and 368 were not present in the reference. The number of resequenced accessions involved in our study was considerably larger than that in the study by Dolatabdian et al. (2020), and consequently, the pangenome was also larger (1789.9 vs. 1040 Mb). We identified 1337 FTRGs in the pangenome, including 616 core and 721 variable genes, indicating high diversity (Fig. 1, Table 2).

In the last two decades, knowledge of FT regulation in model plants has rapidly increased, and analyses of the complicated genetic networks underlying each pathway have revealed multiple connections between the components of these pathways (Berry and Dean 2015). Genetic analysis of FT regulation was greatly facilitated by referring to *Arabidopsis thaliana*, a close relative of the family Brassicaceae. In our study, the *B. napus* FTRGs used for the PAV and SV analyses were identified by aligning *Arabidopsis* FTRGs to the coding sequences of the *B. napus* pangenome. *Arabidopsis* FTRGs were downloaded from the Flowering-Interactive



Fig. 6 Manhattan plots of genome-wide association studies (GWAS) of presence/absence variation (PAV) and structural variation (SV) on flowering time in six environments. The dashed lines represent significance threshold $(-\log_{10}P(PAV)=6.3, -\log_{10}P(SV)=5.7)$

Database (FLOR-ID), a hand-curated database containing information about 306 FTRGs and linked to 1595 publications (Bouché et al. 2016). Due to its polyploid nature, the network of *B. napus* FTRGs comprising approximately 1337 genes was considerably more complicated than that of *Arabidopsis*. The A subgenome of *B. napus* had more variable FTRGs (275 FTRGs) than the C subgenome (227 FTRGs), and the frequencies of PAVs and SVs were higher in the A genome than in the C genome (Tables 2 and 3), indicating that the A genome had higher diversity than the C genome in the FT control. This result was consistent with that of our previous study that compared the overall genetic diversity between the two subgenomes by analyzing SNPs (Wu et al. 2019). The A genome had a higher degree of genetic diversity, possibly due to occasional outcrossing between rapeseed (AACC) and its diploid ancestor species *B. rapa* (AA). In contrast, outcrossing between rapeseed and *B. oleracea* (CC) is very rare. Therefore, it is likely that the genetic diversity of the C genome was limited to a few donors of the original hybridization events that had created the species. In general, rapeseed breeding, especially that of the canola type, is subject to a narrow genetic diversity, and the PAVs and SVs of FTRGs could serve as efficient markers for breeding the ideal FT. Schiessl et al. (2017a; 2017b) identified 184 FTRGs orthologous to 35 *Arabidopsis* FTRGs. The 1337 FTRGs included 145 (78.8%) genes that they identified. The discrepancy might arise from the different thresholds applied to determine orthologous genes, and Fig. 7 Venn diagrams comparing flowering time-related genes determined by a GWAS-SNP, b GWAS-PAV, and c GWAS-SV between different locations. The blue, red, and yellow circles represent XY, HZ, and JX, respectively. The overlapped genes were indicated with numbers, and their IDs are provided in Supplemental Tables S15, S17, and S18. HZ, Hu-Zhou, 30.89°N and 119.63°E; JX, Jia-Xing 30.86°N and 120.70°E; XY, Xian-Yang 34.80°N and 108.10°E (color figure online)



different reference genomes applied. For a part of the FTRGs such as *FLC*, *FT*, *TEMPRANILLO 1 (TEM1)*, and *EARLY FLOWERING 7 (ELF7)*, we found the same number of copies as they did. However, the positions of the copies might be different. For example, they localized one of the two *ELF7* copies on A10 that was the same as we did, but the other *ELF7* copy, which we recognized as *BnaC09G0616900ZS*, on unknown chromosomes. For some other *Arabidopsis* FTRGs such as *AGAMOUS-LIKE 24 (AGL24)*, *APETALA 1 (AP1)*, *CIRCADIAN CLOCK ASSISTED 1 (CCA1)*, *CDF1*, *EARLY FLOWERING IN SHORT DAYS (EFS)*, *FLOWER-ING LOCUS D (FD)*, *FRUITFUL (FUL)*, *GLYCIN-RICH PROTEIN 7 (GRP7)*, *PHYA*, *SRR1*, *VIN3*, we identified more copies than they did. For a few genes such as *CO*, they found more copies.

Based on PAV and SV analyses of FTRGs in *B. napus*, it seems that the spring ecotype and semi-winter ecotype had higher similarity between each other compared with the winter ecotype (Figs. 2, 3). According to Fussell (1955) and

our previous study (Wu et al. 2019), Europe was the center of ancient rapeseed cultivars. *B. napus* could have been cultivated in the Mediterranean areas as early as Classical Times (between the eighth century BC and the sixth century AD). It was spread to other areas by soldiers, merchants and/or birds, and animals, adapting to diverse climate zones and latitudes and forming ecotypes different from the winter type. Winter rapeseed distinguishes from semi-winter and spring types with a longer life cycle and more stringent requirements for vernalization. It is the ancestor type, whereas the semi-winter and spring types were formed in the process of adaptation to eco-environments where seedlings grow in warmer weather conditions.

QTL is a section of DNA that correlates with trait variations such as FT. Mapping of QTL is often an early step in finding genes that are responsible for trait variations. Even with a fine-mapping approach, a QTL interval may cover several candidate genes. In this study, we remapped some known QTL that control FT to the pangenome and analyzed the genetic polymorphisms of the FTRGs underlying 14 known QTL. Mutational variants such as gene loss, stop codon, missense, synonymous, intron, splice regions, and 3'-UTR of the FTRGs underlying the QTL were intuitively displayed (Fig. 4). Mutational variants, such as synonymous variants, can be excluded and PAVs can be prioritized as candidates in the analysis. Moreover, the graph can help in understanding the diversity of a specific FTRG and narrowing the candidate genes that contribute to a QTL effect. Overall, the winter ecotype did not have a higher mutation frequency than the semi-winter and spring ecotypes. However, the deviation of mutation frequency between the individuals within the semi-winter type was obviously bigger than that within the winter and spring types (Fig. 4). This might arise from the choice of the reference genome. In our research, we used ZS11 as the reference, which is a typical native Chinese cultivar. The semi-winter group contain both Chinese cultivars/breeding lines that are genetically close to ZS11, and foreign cultivars/breeding lines from, e.g., Pakistan and/or Australia that may have large genetic distances from ZS11.

Furthermore, to gain more knowledge about the FTRGs and expand the number of FTRG candidates, we performed GWAS-SNP, GWAS-PAV, and GWAS-SV to identify FTRGs after growing a genetic population in three locations in two successive years. XY is located in the northwest region of China (34.80°N and 108.10°E) with a relatively higher latitude where winter is rather cold and dry, and the daily photoperiod in spring is relatively long. HZ (30.89°N and 119.63°E) and JX (30.86°N and 120.70°E) are located at a ~ 100-km distance within the same province Southeast of China, where winter was milder and rainfall was more plentiful than at XY (Fig. S11, Table S19). We observed that the FTO varied across different locations and years. In particular, 24.66% of accessions had a relatively consistent FTR, and 66.10% of accessions had moderate FTR changes. In contrast, 9.25% of accessions had rather drastic FTOV across non-adjacent DSF grades (Fig. 5). To understand whether the FTR change across the different environments depended on ecotypes, we investigated the relationship between the FTR consistency and the ecotypes. As shown in Table 4 and Figure S12, the proportions of each ecotype in the consistent categories and moderately fluctuated categories (Fig. S12d and e) were not much different from those in the whole set of GWAS accessions (the 292 accessions) (Fig. S12c). However, the proportions of the winter ecotype and semi-winter type were over and insufficiently represented, respectively, in the drastically fluctuated category (Fig. S12f), considering their proportions in the whole set. A probable reason is that semi-winter accessions were more adaptable to all three geographic locations, as most of them were native cultivars/lines. But, most winter accessions were from Europe, where the eco-conditions are very

 Table 4
 The consistence of flowering time rank (FTR) across six environments

	The FTR-con- sistent category	The FTR-moder- ately fluctuated category	The FTR-dras- tically fluctu- ated category
Total	72	193	27
Spring	14 (19.44%)	42(21.76%)	6 (22.22%)
Winter	46 (63.89%)	115 (59.59%)	19 (70.37%)
Semi-winter	12 (16.67%)	36 (18.65%)	2 (7.41%)

different from the three experimental locations. They were, therefore, more sensitive to flowering conditions. Overall, the consistency of FTR depends more on individual genotypes. We assume that each flowering pathway might not be equally important for the individuals, and the coordination of homologous genes between the subgenomes of the polyploidy might be unique for a specific genotype. However, we need further experiments to verify this assumption.

GWAS-SNP revealed that JX and HZ shared a total of 152 genes that were associated with FT. In contrast, XY and HZ, XY, and JX only shared 40 and 18 genes that were associated with FT, respectively (Fig. 7a; Table S15), suggesting that different FTRG networks would control FT in varying locations. The greater the distance between two locations, the more different the genes controlling FT. GWAS-PAV and GWAS-SV revealed a significantly lower number of FTassociated genes but did not reveal any FTRGs that were shared by all three locations (Fig. 7b, c). Moreover, relatively fewer associated genes were identified at XY than at HZ and JX (Fig. 7b, c). We enriched the number of FTRGs and provided ID and annotations of the genes (Tables S15–S18). The GWAS-SNP confirmed the importance of FLC in determining FT, which was also reported in our previous study (Wu et al. 2019) and another study (Song et al. 2020). Interestingly, homologous copies of a gene appeared to play different roles in various environments. For example, both BnaA03G0144400ZS and BnaC02G0039100ZS are FLC homologous genes and were identified as loci associated with FT; however, BnaFLC-A03 (BnaA-03G0144400ZS) was identified at HZ, whereas BnaFLC-C02 (BnaC02G0039100ZS) was found at JX. Similarly, both BnaC03G0129200ZS and BnaA02G0118200ZS were VIN3 homologs; however, BnaVIN3-C03 (BnaC03G0129200ZS) was identified at JX, whereas BnaVIN3-A02 (BnaA-02G0118200ZS) was found at XY (Tables S15 and S16). This may be ascribed to epigenetic activation or repression of a locus (such as temporary methylation or demethylation) in a given environment. However, the role of epigenetic modifications in such phenomena requires further investigation. The GWAS revealed a PHYTOCHROME-INTERACT-ING FACTOR 6 (PIF6) ortholog (BnaA09G0562200ZS) at XY, but not at HZ and JX (Tables S15 and S16), implying

that the role of a certain flowering pathway (e.g., the photoperiod pathway) might not be equally important for different locations. The daily sunny hours could be an important factor limiting the transition from the vegetative stage to flowering at XY (Noda and Ozawa 2018). On the other hand, genes in the vernalization pathway could be dominantly important for flowering at HZ and JX, as the winter of the two locations might not have been cold enough to meet the vernalization requirements (Fig. S11).

In conclusion, we provide the highest resolution graphics of FTRGs in *B. napus* so far on a pangenome-wide scale based on SNP and SV analyses. We gain novel insights into the genetic basis of flowering time QTL and the $G \times E$ effect on flowering. We discover that the FTO of plants in a genetic population changes a lot across various environments, and homolog copies of some key FTRGs played different roles in different locations. The results of our study can be of reference in selecting and/or manipulating candidate FTRGs for the introduction and domestication of varieties to a new environment.

Supplementary Information The online version contains supplementary material available at https://doi.org/10.1007/s00122-023-04326-w.

Acknowledgements We thank Dr. Ulrike Lohwasser from Leibniz Institute of Plant Genetics and Crop Plant Research, Germany, for providing a part of the rapeseed accessions for this study, Dr. Li Ruimin from Zhejiang Meteorological Bureau, for providing the meteorological data of the experimental sites, and Mr. Rui Sun from Agricultural Experiment Station of Zhejiang University for the management of field experiments.

Author contribution statement LJ conceived the experiments. YX carried out the field experiments and data analyses. XK contributed a lot to programming and data analyses. YG, RW, XY, XC, YL, JD, YZ, MC, and HC participated in field experiments in multiple locations and years. TY and DW were involved in data analyses and many constructive discussions. YX and LJ wrote the manuscript.

Funding The work was financially sponsored by Natural Science Foundation of China (code nos. 32130076 and 31961143008), Key Science and Technology Project of Zhejiang Province (2021C02057), and Jiangsu Collaborative Innovation Centre for Modern Crop Production.

Data availability The supporting data of Figures and Tables are available in Supplemental Tables S1–S23. The raw reads of the rapeseed accessions have been deposited in the public database of National Center of Biotechnology Information under SRP155312 (https://www.ncbi.nlm.nih.gov /sra/SRP155312) and China National Center for Bioinformation (NGDC) (https://ngdc.cncb.ac.cn/gsa/browse/CRA001854).

Declarations

Competing interests The authors have no relevant financial or non-financial interests to disclose.

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