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**Genome-wide association studies of important agronomic traits in
Brassica napus: what we have learned and where we are headed**

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Abstract: Oilseed rape (*Brassica napus* L.; *B. napus*) is one of the main oil crops in China as well as in the world. Genome-wide association studies (GWAS) have revolutionized the field of complex agronomic traits. In *B. napus*, these include seed yield and yield-related traits, seed oil content, abiotic and biotic stress tolerance traits over the past decade, in which hundreds of thousands to millions of genetic variants across the genomes of hundreds of individuals have been tested to identify genotype-phenotype associations. In this review, we assess the current status of GWAS in terms of genotypes, phenotypes, statistical models and candidate genes for these agronomic traits in *B. napus*. Post-GWAS, the combination of QTL mapping, transcriptomics and new statistical methods has allowed us to identify candidate genes associated with specific agronomic traits. In addition, we can use diverse populations, increase the population size or look for rare variants and structural variations of *B. napus* by whole-genome sequencing to minimize “missing heritability” effects. These approaches are essential for uncovering the genetic mechanisms defining or regulating complex agronomic traits and delivery of molecular marker assisted breeding in *B. napus* to breed new varieties that are higher yielding but resilient to our changing climate.

Key words: Oilseed rape, Agronomic traits, GWAS, Genotypes, Phenotypes, Genetic mechanisms

Introduction

Oilseed rape (OSR, rapeseed, canola; *B. napus*) originated in Mediterranean region approximately 7,500 years ago by natural hybridization between *B. rapa* and *B. oleracea* (Chalhoub et al., 2014; Lu et al., 2019). It has since spread geographically and economically, becoming the second most important oil crop globally for edible oil (USDA ERS, 2020). The growing demand for *B. napus* oil has led to the continuous genetic improvement of various important agronomic traits. However, an increasing or static

yield gap in many countries and our changing climate and reduced efficacy of pesticides means abiotic and biotic pressures are increasing and new resources and approaches to breeding high yielding and resilient germplasm are urgently needed.

Genome-wide association studies (GWAS), in which hundreds of thousands to millions of genetic variants across the genomes of many individuals are tested to identify genotype–phenotype associations (Figure 1), have revolutionized the field of complex agronomic traits of *B. napus* over the past decade (Tang et al., 2021). In 2010 the first GWAS was reported in *B. napus*, in which 684 amplified fragment length polymorphism (AFLP) markers in 84 canola quality winter rapeseed cultivars were used to explore the genes affecting phenological, morphological and quality traits (Honsdorf et al., 2010). To date, diverse traits of *B. napus*, ranging from molecular scale (e.g. the transcriptome) to whole plant scale (e.g. seed yield, seed oil content (SOC), root and shoot architecture traits, flowering time), and from abiotic stress tolerance (e.g. in response to phosphorus (P) deficiency, salt and drought stress etc.) to biotic stress tolerance (e.g. stem rot and clubroot), have been studied using GWAS approaches (Table 1). The number of articles on GWAS in *B. napus* has increased year by year (Figure 2), which suggests that this approach is a powerful tool to analyze the genetic structure of important agronomic traits in *B. napus*. Herein, we (1) review the advances on the *B. napus* functional genomics facilitated by GWAS; and (2) assess the challenges and opportunities of *B. napus* GWAS going forward.

Genotypes used in GWAS in B. napus

Single nucleotide polymorphism (SNP) chips and whole- genome sequencing (WGS) are the most widely used genotyping methods currently used to identify genetic variants within genotypes from populations used for GWAS. To date, the 60k SNP chips have been used in approximately 54% of *B. napus* GWAS (Table 1). However, with the development of WGS technology, GWAS based on high-

depth WGS is gradually increasing (Wu et al., 2019; Lu et al., 2019; Tang et al., 2021). Compared with SNP chips, WGS is capable of detecting more genetic variants, especially rare and structural variations, but the cost is higher, and the analysis method is more complicated (Table 2).

GWAS is primarily designed to test SNVs (single-nucleotide variants) for association with traits. However, other types of genetic variants can also be detected by GWAS. For example, structural variations (SVs), including small insertions and deletions (InDels) and presence–absence variation (PAV), were prevalent in the *B. napus* genome and some studies show they make significant contributions to *B. napus* phenotypic variations (Gabur et al., 2018; Song et al., 2020). A hundred and fifteen significant PAV were associated with blackleg resistance in *B. napus*, of which, 41 were significant at false discovery rate (FDR) ≤ 0.1 (Gabur et al., 2018). Recently, more than one million PAVs were identified and enriched in associations with silique length, seed weight and flowering time of *B. napus* by performing *de novo* assembly and annotation of eight representative *B. napus* genomes (Song et al., 2020). Compared to SNPs, PAVs contribute much more to these phenotypes (Song et al., 2020).

Besides SNVs and SVs, transcriptional variation is also widely used in GWAS of *B. napus* (Harper et al., 2012; Lu et al 2014; Alcock et al., 2017; Alcock et al., 2018; Tang et al., 2021). GWAS for transcriptomic variation is also called transcriptome-wide association studies (TWAS) or associative transcriptomics. GWAS with both SNPs and gene expression markers (GEMs) identified *Bna.HACG*, a gene that controls glucosinolate biosynthesis in *B. napus* seeds (Harper et al., 2012). GWAS combined with TWAS revealed two candidate genes, *Bna.A10.ACA8* and *Bna.C02.MGT7*, responsible for the leaf accumulation of calcium and magnesium, respectively (Alcock et al., 2017). Subsequently, a total of five GEMs markers associated with leaf potassium concentration were identified, and 9 candidate genes were predicted (Alcock et al., 2018). Recently, a combination of GWAS and TWAS was employed to uncover

the regulatory networks associated with SOC in *B. napus* and each expression quantitative trait loci (eQTL) can largely explain the variation of gene expression signatures (Tang et al., 2021).

Phenotypes used in GWAS in B. napus

Natural population is the most commonly used population for GWAS analysis in *B. napus*, and generally consists of three types of rapeseed, i.e. spring, winter and semi-winter (Wu et al., 2019; Lu et al., 2019; Tang et al., 2021). However, in some studies, a few *B. napus* types within the experimental population were probably not adapted to the climate of the growing area. For example, spring type rapeseed is not suitable for the cold climate because it will cause low seed yield, but in some GWAS, the spring, winter, and semi-winter rapeseed were planted in the same growing area without considering the impact of climate to study the yield and yield-related traits (Sun et al., 2016b; Lu et al., 2017; Li et al., 2021). Consequently, the differences in the seed yield and yield related traits are mostly attributed to the growing environment, rather than the yield-related genetic factors. Therefore, researchers should construct a reasonable experimental population according to the adaptability (e.g. climate) of the plants. Some researchers have also noticed this problem and made a reasonable design for the experimental population (Kumar et al., 2018). In France, the spring type rapeseed does not grow well, which has resulted in a compromised investigation into the impact of blackleg disease. Considering this, a population of 166 winter *B. napus* were planted in France to conduct GWAS analysis of blackleg disease (Kumar et al., 2018).

In *B. napus* GWAS, some phenotypic traits, such as flowering time, seed yield, plant height, branch number, and seed weight, can be directly measured (Figure 3a); some other traits such as tolerance to biotic and abiotic stresses, cannot be directly investigated. For the latter, some parameters, such as root length under P deficiency (Wang et al., 2017b), shoot fresh weight under salt stress (Wan et al., 2017),

root to shoot length ratio under water stress (Zhang et al., 2015), radicle length under lead (Pb) stress
 (Zhang et al., 2020), seed vigor under low-temperature stress (Luo et al., 2021), the stem rot disease
 index (Wei et al., 2016) and root fresh weight under drought stress (Khanzada et al., 2020) were used to
 conduct GWAS. Although these parameters could indicate the ability of plants to tolerate stress they
 increase the complexity and the difficulty of uncovering the genetic mechanism in response to abiotic
 and biotic stress by GWAS. To overcome this, one opportunity is to quantify the metabolites directly
 related to the stress in *B. napus* and perform metabolite-based GWAS (Figure 3b). Metabolite-based
 GWAS has previously been used to identify 1,459 significant locus–trait associations in maize, with more
 than 58.5% identified as significant loci for 983 metabolites in maize kernel supported by expression
 QTLs, and some (14.7%) co-located with QTLs in previous linkage analysis, and two genes were further
 validated by mutant and transgenic analysis (Wen et al., 2014). Metabolite association studies will be
 helpful to understand the functional mechanism of complex traits and identify the candidate gene(s).
 GWAS of metabolites of *B. napus* under abiotic and biotic stress can be performed in the future.

Four strategies are useful to improve the precision of an experiment: (1) Blocking, (2) Randomization,
 (3) Replication, and (4) Repeat measurements. The first two are smart strategies because they do not
 require large samples. The last two are costly strategies, but they are necessary if random and
 measurement errors must be estimated and controlled, such as one-year/multi-location experiments, one-
 location/multi-year experiments, multi-year / multi-location experiments. Best linear unbiased prediction
 (BLUP) analysis has been used successfully to deal with multi-year phenotypic values of agronomic
 traits, such as plant height and branch number (Liu et al., 2021b), silique number (Li et al., 2020), seed
 oil content (Wang et al., 2018), seed phytate content (Liu, et al., 2021a) in rapeseed, and ionic traits
 in rice (Yang et al., 2018). R package ‘lme4’ (<https://cran.r-project.org/web/packages/lme4/index.html>)

is used to calculate the BLUP values of investigated traits.

GWAS of important agronomic traits in B. napus

GWAS of seed yield and seed yield related traits, including plant height, first branch height, inflorescence length, silique length, seeds per silique and seed weight of *B. napus* were first reported in 2014 (Cai et al., 2014; Table 1). A panel of 192 inbred lines of *B. napus* were genotyped using 740 AFLP markers, and a total of 43 AFLP markers were identified associated with yield and yield-related traits (Cai et al., 2014; Table 1). Five common markers were associated with both first branch height and plant height, and one marker was associated with both silique length and seed weight in the three-year experiment (Cai et al., 2014; Table 1). In a larger study, using a panel of 157 *B. napus* accessions and 690953 SNPs, 20 SNPs were significantly associated with seed weight on A01, A04, A09, C02, and C06 chromosomes (Dong et al., 2018; Table 1). Recently, six candidate genes for silique number were identified through a combination of GWAS and transcriptome profiling (Li et al., 2020a; Table 1). This study also highlighted that natural variation of silique number is largely affected by *B. napus* biomass and nutrient accumulation, which highlighted a potential new approach for the genetic improvement of silique number in *B. napus* (Li et al., 2020a). Recently, GWAS combined with QTL analysis revealed the genetic architecture of seed yield at high planting density in *B. napus* and identified two candidate genes (*BnaA02.TCPI* and *BnaA02.HY5*) for it (Menendez et al., 2021). Overexpression of *BnaA02.TCPI* in *Arabidopsis thaliana* promoted growth independently of plant density, and overexpression of *BnaA02.HY5* increased seed yield and biomass at high planting density (Menendez et al., 2021).

One hundred and forty three SNPs were significantly associated with flowering time by a GWAS study with 368 *B. napus* accessions and 11,804 SNPs, and *Bna.A02.FT* and *Bna.C02.FT* were identified as candidate genes (Raman et al., 2019). In another GWAS study for flowering time with 991 *B. napus* accessions and more than nine million SNPs, *Bna.A02.FT* was also detected (Wu et al., 2019).

Subsequently, *Bna.A02.FLC* and *Bna.A02.FT* were confirmed as candidate genes for flowering time by GWAS study of flowering time with a panel of 950 F1 hybrids from 475 spring-type canola (Jan et al., 2019). In these studies, *Bna.A02.FT* was mapped by different research groups using different *B. napus* accessions, showing its dominant effect in controlling the variation of flowering time in *B. napus*.

For SOC, 39 SNPs were identified by a combined GWAS (227 *B. napus* accessions and 34,000 high-quality SNP markers) and a QTL mapping approach, with explained phenotypic variations ranging from 10.23% to 24.45%. SNPs located on A05, A08 and A10 co-located with previously reported QTLs associated with SOC (Sun et al., 2016c). Herein, the QTL means the significant SNPs with close proximity (Sun et al., 2016c; Li et al., 2021). In another GWAS study with 521 *B. napus* cultivars and 52157 SNPs, 50 SNPs were significantly associated with SOC, of which, 29 were newly discovered (Liu et al., 2016b). Recently, 27 loci for SOC were identified, and two candidate genes of *BnaA05.PMT6* and *BnaC05.PMT6* were confirmed as negative regulators of SOC by a combined GWAS and TWAS using a panel of 505 *B. napus* accessions and more than 10 million SNPs from WGS (Tang et al., 2021). This is the first study that used GWAS/TWAS to identify and confirm candidate genes in *B. napus* using knockout mutants and overexpression lines.

GWAS of root related traits under P deficiency identified 285 SNP loci associated with these traits, and a haplotype of '*BnA03Hap*' on A03 chromosome, which will be important for breeding high P efficient varieties (Wang et al., 2017b). Nine SNPs associated with Pb accumulation in *B. napus* were identified by GWAS and the expression of five candidate genes (*GSTUs*, *BCATs*, *UBP13*, *TBR* and *HIPP01*) were verified by Quantitative Real-time PCR (qRT-PCR) (Zhang et al., 2020). GWAS was undertaken to uncover the genetic basis of low-temperature stress in *B. napus* using more than 8 million SNPs by WGS, and 22 loci were identified. Underlying these loci, five candidate genes, *BnaA03g40290D*,

BnaA06g07530D, *BnaA09g06240D*, *BnaA09g06250D*, and *BnaC02g10720D* associated with low-temperature stress were confirmed (Luo et al., 2021).

In biotic interaction, GWAS identified 17 significant SNPs associated with stem rot resistance on chromosomes A08 and C06 by using an association panel with 347 *B. napus* accessions, with a *tau class glutathione S-transferase* gene cluster identified by combining GWAS and transcriptome analysis (Wei et al., 2016). Eight co-located SNPs for blackleg (*Leptosphaeria maculans*) resistance in *B. napus* were identified by combining GWAS analysis and QTL mapping (Raman et al., 2016). Subsequently, sixteen significant SNPs associated with blackleg disease were identified by combining GWAS analysis and QTL mapping, of which nine were co-located with the genomic regions in a previous study (Kumar et al., 2018).

In total, thousands of genetic variations have been identified for various traits in *B. napus* by GWAS in recent years. If the genetic variations associated with seed yield and yield-related traits are also associated with other traits, such as biotic and abiotic stress resistance traits, there may be a gene controlling both traits (pleiotropy) near the location of genetic variation.

A tip of the iceberg – current GWAS in B. napus

Interplay of genetics and environmental factors on complex agronomic traits

Complex agronomic traits result from the interplay of genetics and environmental factors. For example, seed yield in *B. napus* is dependent on complex interactions between genetic predisposition, abiotic stress (e.g. water or nutrients deficiency) and biotic stress (e.g. *Sclerotinia Sclerotiorum*, Clubroot Disease) (Zhang et al., 2015; Wang et al., 2017b; Fredua-Agyeman et al., 2020). For example, P is an essential nutrient in the regulation of plant growth and development and P deficiency reduces seed yield significantly. However, in most of the GWAS studies on the seed yield of *B. napus*, soil nutrient

availability is not reported. Given there are significant differences in the P use efficiency among different varieties in an association panel of *B. napus* (Wang et al., 2017b) or if the soil used for phenotyping is severely P deficient, the seed yield trait will contain a component of the adaptation to deficient P conditions, and the genes associated with seed yield may also be associated with the P use efficiency. With most GWAS only focusing on specific phenotypes, even minimizing environmental variables, there is the possibility that the measured trait and SNP associations are the product of these interactions. In addition, different growth stages, different planting location, and the impact of abiotic and biotic stresses may affect the final yield in different ways and through different genetic and metabolic interactions (Figure 4). GWAS findings today therefore represent the starting point for the exploration of a wider range of phenotype-environment interactions through GWAS and is likely to lead to additional discoveries (Figure 4).

Populations and population size

Most populations used in GWAS of *B. napus* are natural populations (Table 1). The abundant diversity of natural populations can make GWAS reach single variance level resolution. However, the population structure, kinship and rare variants in natural populations may affect the statistical power of GWAS (Flint-Garcia et al., 2005). It is difficult to detect rare variants underlying target traits if they are significantly affected by the population structure (Flint-Garcia et al., 2005). To overcome false associations and improve detection ability of rare alleles in crops, multi-parent cross populations have been developed, including nested association mapping (NAM), random-open-parent association mapping (ROAM), and multi-parent advanced generation Inter-Crosses (MAGIC) (Yu and Buckler, 2006; Dell'Acqua et al., 2015; Xiao et al., 2017). Recently, a *B. napus* NAM population consisting of 2425 F₆ RILs derived from crosses between the common parent Zhongshuang11 (ZS11) and 15 diverse semi-

winter oilseed rape and spring oilseed rape founder lines, has been developed to dissect the genetic architecture of some important agronomic traits, such as silique length, seed weight and flowering time (Hu et al., 2018; Song et al., 2020). In future research, the complementary advantages of natural and artificial populations will further accelerate the dissection of important agronomic traits in *B. napus* (Figure 4). In addition, larger sample sizes in association populations are necessary for the identification of additional loci and rare variants because sample size is the primary limiting factor in genetic variation discovery. A population of 200 individuals can detect QTLs that explain 20% of the phenotypic variance with 90% power. However, if the QTLs only explained 5% of the phenotypic variation, the population would need to have more than 1000 individuals with the same power of detection (Gatti et al., 2014). Therefore, using GWAS to identify a new trait, where its phenotypic variance is not clear, and increasing the population size could improve the detection efficiency of the phenotypic variance (Figure 4).

Statistical models

Statistical models are an important part hidden in the GWAS “iceberg”. GLM and MLM are two models commonly used in *B. napus* GWAS (Figure 4, Table 1). GLM only uses the population structure, and the outputs have a high level of false positives. MLM uses both the population structure and kinship; however, the calculation speed is slower and the computational resources are higher compared with GLM. In recent years, large-scale whole-genome resequencing of *B. napus* natural populations had been carried out (Wu et al., 2018; Lu et al., 2019; Tang et al., 2021). Several models, such as EMMXA (Wu et al., 2018;), mrMLM (Lu et al., 2019; Qian et al., 2021), GEMMA (Luo et al., 2021) and Fast-LMM (Tang et al., 2021), have been used to conduct GWAS based on WGS data, which have faster calculation speeds and higher powers of detection. In addition, a method has been developed to predict candidate genes in each QTL based on multi-layered data and information of *Arabidopsis* (Tang et al., 2021). The use and

development of new statistical models will greatly improve the identification of candidate genes in the future (Figure 4).

Rare variation and structural variants

GWAS has been used to identify thousands genes for human diseases and agronomic traits in crops, which is accelerating our understanding of the genetic components of these traits, such as age-related macular degeneration in human, grain weight in rice and SOC in oilseed rape (Klein et al., 2005; Si et al., 2016; Tang et al., 2021). However, only a small portion of phenotypic variation can be explained in almost all the studied traits using GWAS, especially in humans (Génin et al., 2020). For example, across 5000 human GWAS, 68,000 SNP-trait associations were identified in humans (the GWAS catalog, April 2018). These SNPs only explain a modest proportion of the estimated heritability of all the traits except for those such as age-related macular degeneration and exfoliation glaucoma (Manolio et al., 2009; Manolio et al., 2013; Buniello et al. 2019). This is known as the “missing heritability”. For example, in *B. napus*, 285 root related SNPs have been identified, but only 5.8% of them explained 10% of the phenotypic variation (Wang et al., 2017b). The first reason for this is GWAS is not capable of capturing all the genetic associations with target traits, and most of the variants with small effects are missed. The second reason is that the rare variations which are not captured by SNP-chip based GWAS may be the main factor(s) controlling the traits. For example, the final genotype files used for SNP-chip based GWAS in *B. napus* do not contain rare variants because the genotype files are usually filtered according to minor-allele frequency (MAF) > 0.05, which may lead to the rare variants that are associated with the phenotype not being represented and identified. In addition, structural variants are not represented in SNP chips, which are completely ignored in early GWAS analysis of complex traits. Structural variants are important genome variations in *B. napus*, which encompass more nucleotides on the genome than SNPs (Song et

al., 2020). Incorporation and analysis of structural variants-GWAS are proposed to promote the genetic analysis of complex agronomic traits in *B. napus* in the future.

Connecting GWAS to biological mechanisms underlying complex traits

Although most of *B. napus* GWAS have focused on candidate gene mining, few new genes have been cloned or validated through functional molecular approaches. This could be attributed to the limitations of GWAS, such as target traits being associated with unrelated loci if they are significantly correlated to population structure and kinship, which is a common cause of false positives (Flint-Garcia et al., 2005; Larsson et al., 2013). The values of LD decay of *B. napus* are associated with the population size, such as 298 kb in 300 inbred lines (Zhou et al., 2018), 238 kb in 505 inbred lines (Liu et al., 2021a), 0.7 Mb in 280 inbred lines (Li et al., 2021), 1.6 Mb in 177 inbred lines (Dakouri et al., 2021). Nevertheless, the LD decays observed in *B. napus* are larger than that of other species, e.g. 50 kb in *Arabidopsis* (Nordborg et al., 2005), 200 kb in rice (Li et al., 2017) and 160 kb in maize (Li et al., 2019). Large LD decay in *B. napus* indicates an association locus contains more than 100 genes, and thus it is difficult to pinpoint the causal genes associated with the significant loci.

Consequently, loci identified by GWAS associated with traits may only explain a small fraction of the heritability of the target traits and may represent a false association or not necessarily causal variants and candidate genes (Tam et al., 2019). These shortcomings highlight the need to be cautious in the interpretation of GWAS (Tam et al., 2019). Here, a method to narrow down the range of candidate genes are summarized. Firstly, GWAS can be combined with other methods, such as transcriptomics, to predict the candidate genes. Twenty seven percent of the published papers on GWAS of *B. napus* indicate that the differentially expressed genes in the target interval are causal genes (Table 1). Secondly, candidate gene association analysis, especially the differentially expressed candidate genes, are performed to

identify whether there are SNPs significantly related to the target phenotype. If the SNP is significantly associated with the phenotype, its position within the candidate gene (e.g. the promoter, 5'UTR, intron or exon) and its subsequent impact on expression level or protein sequence should be explored. For example, there are two genes (*BnaA10g09290D*, *BnaC08g26640D*) identified by GWAS as candidate genes for controlling plant height under low phosphorus stress (Liu et al., 2021b). Candidate gene association analysis demonstrated that chrA10_8216680 (T/A), located in the exon region of the candidate gene *BnaA10g09290D*, resulted in amino acid changes from isoleucine to asparagine, and the SNP of chrC08_27999778 (A/T) located in the exon region of the candidate gene *BnaC08g26640D* resulted in amino acid changes from isoleucine to asparagine. Both SNPs showed strong association with the trait; therefore, these two SNPs should be the direct effect sites underlying the loci that contribute to the phenotypic difference in plant height under low phosphorus stress (Liu et al., 2021b). In short, if the candidate gene harbors a SNP that causes a significant differentiation in expression levels or protein sequence and that also shows strong association with the target phenotype, the gene is most likely to be the causal gene.

In addition, GWAS researchers should be encouraged to take on the challenge and go beyond identification of loci, with the functional validation, through molecular approaches of causal candidate genes (Figure 4). Recently, 424 significant SNPs were associated with trichome appearance in a *B. napus* GWAS, and a candidate gene, *BnaA06.GLI.a*, was identified by the combination of GWAS with RNA-seq. Functional validation using *Arabidopsis* lines carrying the 35S:*BnaA.GLI.a* cassette formed trichomes on the stems and leaves, confirming its role in trichome development and indicating that *B. napus* might have a similar regulatory network that controls the development of trichome to *Arabidopsis* (Xuan et al., 2020). Twenty-seven loci for SOC were detected and two candidate genes *BnaA05.PMT6*

and *BnaC05.PMT6* were identified. In the knockout mutants of *BnaA05.PMT6* and *BnaC05.PMT6*, the SOC of *B. napus* increased by 3.0% (Tang et al., 2021). Recently, GWAS analysis combined with the transcriptome identified the candidate gene *BnaA07.FAXI-1* responsible for the natural variation of total plant dry matter. The *B. napus BnaA07. FAXI-1* overexpression lines grown under field conditions showed significantly higher biological yield, seed yield, and seed oil content compared with the wild type (Xiao et al., 2021). Fifteen SNPs were identified to be associated with the seed glucosinolates by GWAS and four candidate genes were predicted by GWAS combined with TWAS, comprehensive bioinformatics, and pocket algorithm analyses. Among these four genes, knocking-out only the *B. napus BnaC02.GTR2* resulted in lower glucosinolates acid and higher oil content in seeds, and therefore this gene is likely to be the causal gene (Tan et al., 2021).

The home of GWAS: From laboratory to field

To highlight the power of GWAS, a successful case from laboratory to field was demonstrated in the development of maize varieties with increased concentrations of vitamin A (VA). The candidate alleles of *LcyE* and *crtRBI* were detected by a candidate gene association study (Harjes et al., 2008), and then were introgressed into high yielding maize varieties by molecular marker assisted breeding. The maize with high VA concentrations is now being consumed by malnourished children in Africa, with the prevalence of VA deficiency declining in this region (Fiedler et al., 2014). *B. napus* GWAS have detected many favorable alleles and haplotypes, which can be used for further molecular marker-assisted breeding. For example, eight successive SNPs located near the peak SNP Bn-A03-p3052882 on A03 of *B. napus* formed nine haplotypes, designated '*BnA03Hap*' haplotypes, and the inbred lines carrying GAAAAAGG at '*BnA03Hap6*' have higher than average values in primary root length, total root length, lateral root length, and root dry weight than those with the other haplotypes (Wang et al., 2017b). Recently, GWAS

of plant height of *B. napus* at low P supply identified 133 SNPs for both plant height and branch number and proposed three candidate genes (*BnaA10g09290D*, *BnaC08g26640D* and *BnaA02g33340D*) for plant height (Liu et al., 2021b). Candidate gene association and haplotype analysis identified several P-efficient haplotypes, e.g. *BnaA10g09290Hap1* and *BnaC08g26640Hap1*, which have more branch number and higher plant height than P-inefficiency haplotypes (Liu et al., 2021b). One hundred and nineteen significant SNPs were identified to be associated with phytate concentration in *B. napus*, and a low phytate haplotype (*'BnaA09g10220ConcHap1'*) of the candidate gene (*BnaA9.MRP5*) was proposed (Liu et al., 2021a). In the future, genomic region-based or gene-based GWAS will provide greater confidence in associations between loci and traits, and will contribute to the subsequent breeding of higher yielding and more resilient varieties of *B. napus* (Figure 4).

Conclusions

The emergence of GWAS has caused a remarkable shift in our capacity to understand the genetic basis of plant agronomic and economic traits. At present, a large number of papers have reported the application of GWAS in the analysis of various agronomic traits of *B. napus*. These studies have identified large-scale genetic variation in seed yield and yield related traits, SOC, root architecture traits in response to P deficiency and established the reliability of association analysis. The combination of QTL mapping, transcriptomics, metabolomics or other methods may help to narrow down the candidate genes. In addition, using more diverse populations, increasing the population size or identifying rare or structural variations of *B. napus* by WGS to minimize “missing heritability” effects, will help to reveal more of the ‘GWAS discoveries’ iceberg.

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Conflict of interest

The authors declare no competing financial interest.

Author contributions

HJL, PJW and LS developed the concept. HJL performed data analysis. HJL, JH and LS wrote the manuscript. WW, MY, PY, GJK, GDD, SLW, HMC, CW, CGL and FSX revised the manuscript.

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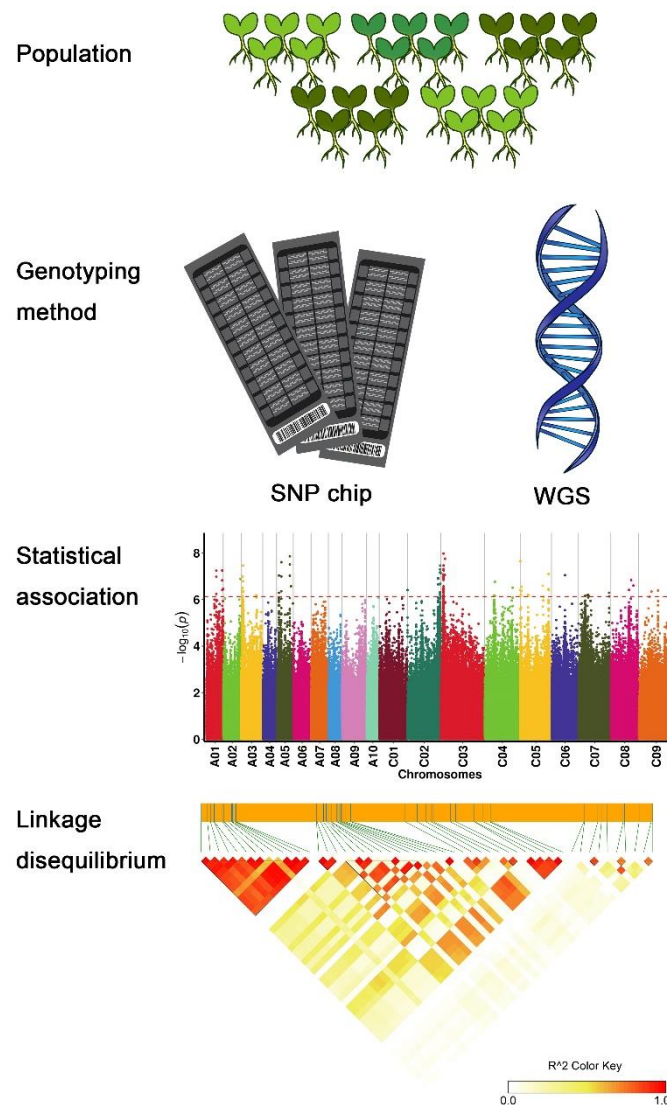
Table 1. GWAS of important agronomic traits in *B. napus* and the candidate genes

(Please see Suppl. file)

Table 2. Benefits and limitations using SNP chip and WGS in GWAS analysis of *B. napus*.

Factors	SNP chip	WGS
Cost	Relatively inexpensive	Expensive (>US\$200 per sample)
Reliability	Reliable, very comprehensive solutions	Less comprehensive solution and less accurate technology
Genomic coverage	About tens of thousands of SNP markers, and the location of SNP markers is fixed, new mutations, and rare mutations and indel markers are not included	Covering the whole genome, from SNP markers, rare variation to almost all genetic variation in the genome
GWAS analysis	Relatively simple, with comprehensive solutions, and can be completed in Windows system	Relatively difficult; requires Linux system and R language
Computational costs	Relatively low; personal computers can meet the needs of computing	Greater costs to store, process, analyse and interpret the resulting data
Research objectives	(1) New traits that have not been studied; (2) Detection of common-variant associations in a large sample size	(1) Traits that have been studied by SNP chip; (2) rare and structure variants

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Fig 1. GWAS study design in *B. napus*. The aim of GWAS is to detect associations between genotype frequency and trait status. The first step of GWAS involves in collecting natural populations and identifying the status of the studied traits. Genotyping can be performed using SNP-chip or WGS. Association analysis is used to identify regions of the genome associated with the target traits at genome- wide. Candidate genes are usually not directly identified but are in linkage disequilibrium with the significant SNPs. GWAS, genome- wide association study; WGS, whole-genome sequencing; SNP, single nucleotide polymorphism.

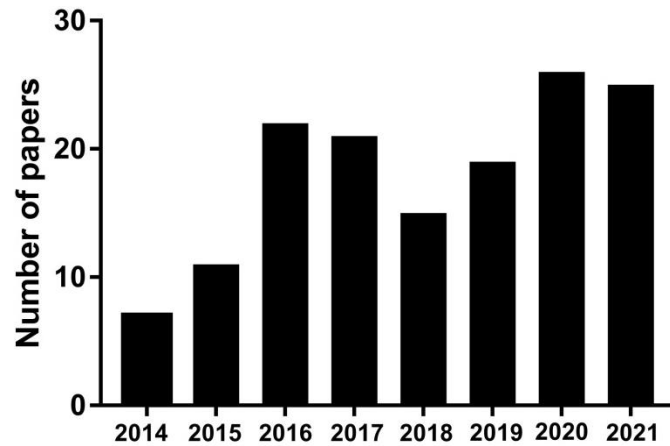


Fig 2. The increase of publications on the GWAS of *B. napus* since the release of the reference genome of *B. napus* cultivar Darmor-bzh in 2014. The data are from NCBI pubmed (<https://pubmed.ncbi.nlm.nih.gov/>). The key words of "genome-wide association study" and "*Brassica napus*" were searched in September 30, 2021.

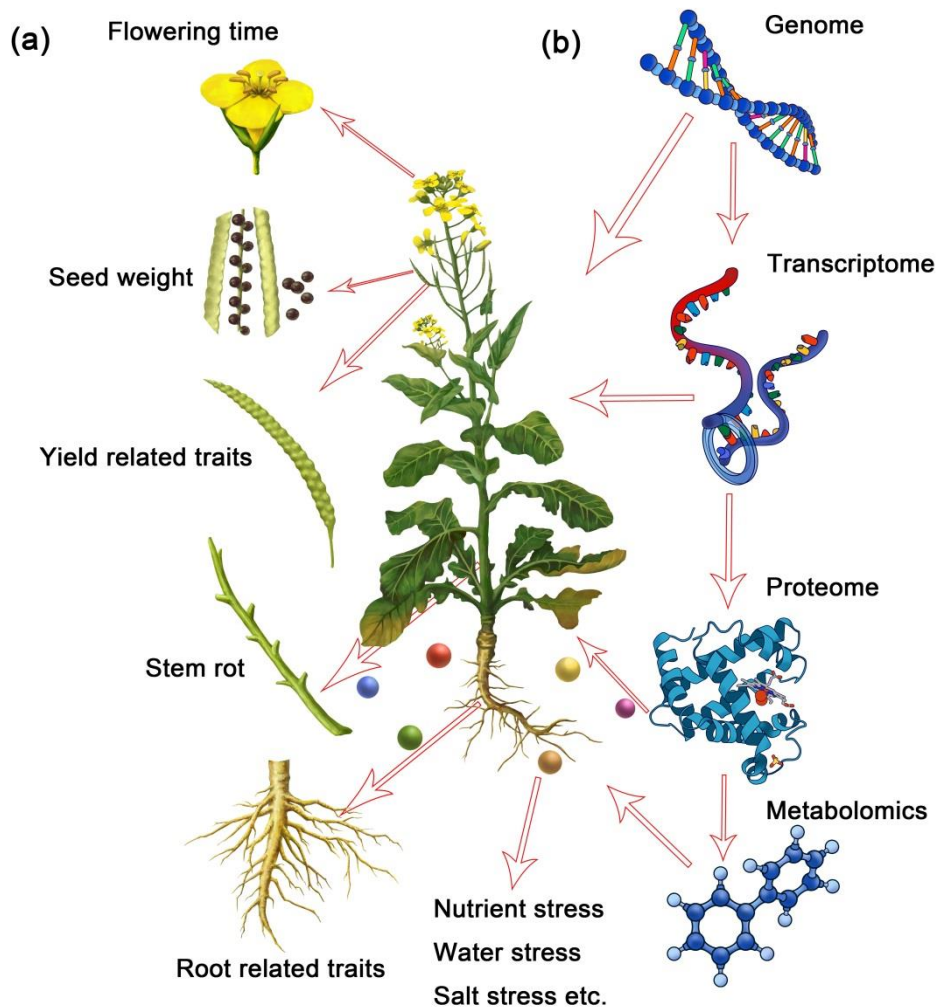


Fig 3. The traits of *B. napus* used in the GWAS. (a) The important agronomic traits today. (b) The omics traits tomorrow.

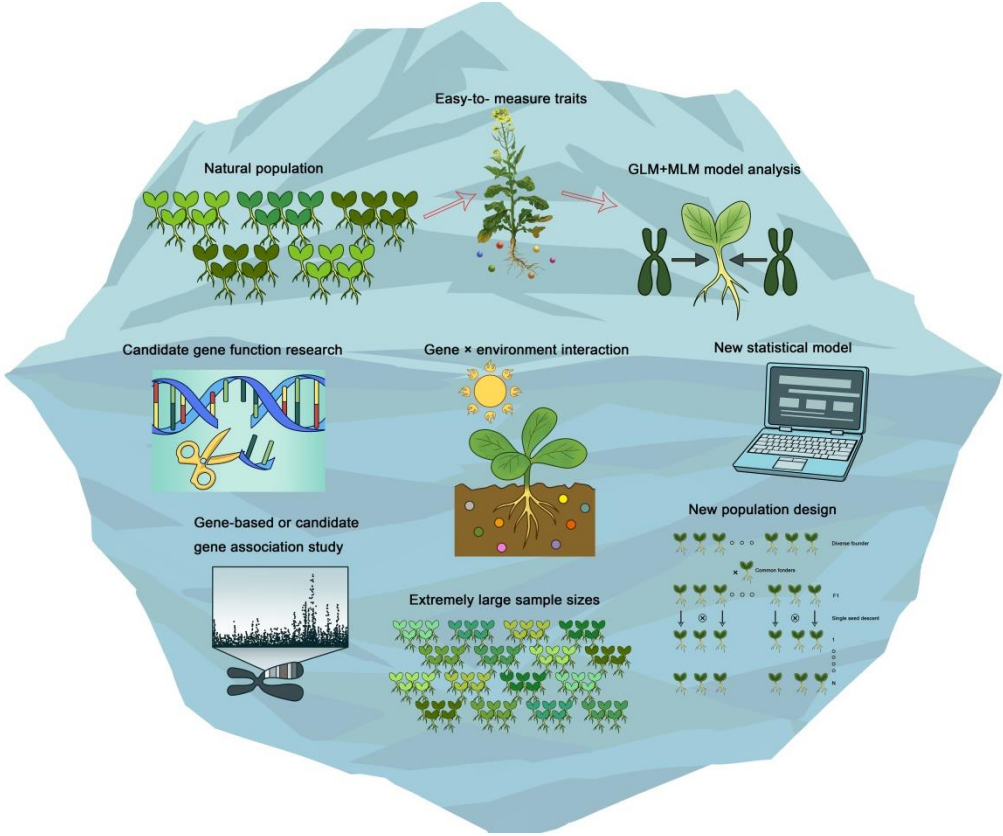


Fig 4. A tip of the iceberg - current GWAS in *B. napus*. The discoveries that can be made in *B. napus* using GWAS are represented by an iceberg. The portion of the iceberg above water represents the discoveries that have been made in *B. napus* by GWAS to date, using easy- to-measure traits (for example seed yield, seed weight and flowering time, etc.), predominantly natural populations; and GLM and MLM models. Most of the iceberg is submerged. The submerged part includes the main exploration directions in the future, including the interaction between gene and environment, larger population sizes, new statistical model, new population design, gene function study, and molecular marker assisted breeding.