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Validation of an updated Associative Transcriptomics platform for the polyploid crop species *Brassica napus* by dissection of the genetic architecture of erucic acid and tocopherol isoform variation in seeds

Lenka Havlickova¹, Zhesi He¹, Lihong Wang¹, Swen Langer¹, Andrea L. Harper¹, Harjeevan Kaur¹, Martin R. Broadley³, Vasilis Gegas² and Ian Bancroft¹*

¹ Department of Biology, University of York, Heslington, York, YO10 5DD, UK
² Limagrain, Joseph Nickerson Research Centre, Rothwell, LN7 6DT, UK
³ Plant and Crop Sciences Division, School of Biosciences, University of Nottingham, Sutton Bonington Campus, Loughborough LE12 5RD, UK

*Correspondence to:
Prof. Ian Bancroft, Department of Biology, University of York, Heslington, York, YO10 5DD, UK. Email: ian.bancroft@york.ac.uk Tel: +44 (0)1904 328778

Author email addresses:
Lenka Havlickova lenka.havlickova@york.ac.uk
Zhesi He zhesi.he@york.ac.uk
Lihong Wang sophia.cheng@york.ac.uk
Swen Langer swen.langer@york.ac.uk
Andrea L. Harper andrea.harper@york.ac.uk
Harjeevan Kaur hk701@york.ac.uk
Martin R. Broadley martin.broadley@nottingham.ac.uk
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**Summary**

An updated platform was developed to underpin association genetics studies in the polyploid crop species *Brassica napus*. Based on $1.92 \times 10^{12}$ bases of leaf mRNAseq data, functional genotypes, comprising 355,536 single nucleotide polymorphism markers and transcript abundance were scored across a 383-accession genetic diversity panel using a transcriptome reference comprising 116,098 ordered CDS gene models. The use of the platform for Associative Transcriptomics was first tested by analysing the genetic architecture of variation for seed erucic acid content, as high erucic rapeseed oil is highly valued for a variety of applications in industry. Known loci were identified, along with a previously undetected minor effect locus. The platform was then used to analyse variation for the relative proportions of tocopherol (Vitamin E) forms in seeds and the validity of the most significant markers assessed using a take-one-out approach. Furthermore, the analysis implicated expression variation of the gene Bo2g050970.1, an orthologue of VTE4 (which encodes a γ-tocopherol methyl transferase converting γ-tocopherol into α-tocopherol), associated with the observed trait variation. The establishment of the first full-scale Associative Transcriptomics platform for *B. napus* enables rapid progress to be made towards an understanding of the genetic architecture of trait variation in this important species and provides an exemplar for other crops.

**Significance statement**


The availability of a full-scale association genetics platform for *Brassica napus*, based on functional genotypes, enables the genetic architecture of essentially any trait to be addressed in this important crop species.

**Introduction**

As the demand for ever increasing crop productivity continues against the backdrop of changing climate and diminishing resources, crop improvement has become an important driver for advances in genomic technologies in plants. A broad aim of crop science is the identification of the genetic bases for trait variation, including both the identification of beneficial alleles and the development of molecular markers to accelerate introduction into elite germplasm. Genetic diversity panels, typically comprising past and current cultivars along with wild relatives, are usually available for crop species. Such panels represent ideal resources for genome-wide association studies (GWAS), which exploit historical recombination between molecular markers and loci associated with trait variation. Where recombination between loci is observed proportionately less frequently than expected for unlinked loci (i.e. < 0.5), those loci are said to be in Linkage Disequilibrium (LD). The approach of identifying molecular markers in LD with loci associated with trait variation is an important tool in human genetics studies and has been applied successfully in several plant species (Garrigan and Hammer, 2006; Li et al., 2008; Atwell et al., 2010; Cockram et al., 2010; Tian et al., 2011; Zhao et al., 2011). The recent development of transcriptome-based GWAS, including the technology termed Associative
Transcriptomics (AT), in which both gene sequence variation and transcript abundance variation are used to identify associations with trait variation (Harper et al., 2012) greatly increases the range of crops to which GWAS approaches can be applied.

The Brassicaceae family includes *Arabidopsis thaliana*, the first plant for which a high quality genome sequence was available (AGI, 2000), and the *Brassica* crops. The diploid species *Brassica rapa* and *Brassica oleracea*, which contain the *Brassica* A and C genomes, respectively, are closely related, having shared a common ancestor only ca. 3.7 Mya (Inaba and Nishio, 2002). *Brassica napus* is an allopolyploid, arising from the hybridization of these species (U, 1935) and the related (homoeologous) regions of the genomes are clearly discernible (Bancroft et al., 2015). A diverse range of *B. napus* crop types have been developed, including oilseed rape, fodders, leafy vegetables and root vegetables. *Brassica* species have been used extensively in genomics studies, due to their utility in studying the evolution of polyploid genomes (Song et al., 1995; O'Neill and Bancroft, 2000; Pires et al., 2004; Yang et al., 2006; Town et al., 2006; Cheung et al., 2009). A draft genome sequence has been obtained for *B. napus* (Chalhoub et al., 2014). However, at ca. 1.2 Gb, the genome of *B. napus* is relatively large. To address this problem, rapid and cost-effective transcriptome-based technologies, using mRNAseq, have been developed and applied for SNP discovery (Trick et al., 2009), linkage mapping and genome characterization (Bancroft et al., 2011) and transcript quantification (Higgins et al., 2012). Indeed, AT was first developed in *B. napus* with a very small genetic diversity panel,
enabling the implication of orthologues of HAG1 in the control of seed glucosinolate content (Harper et al., 2012).

Vegetable oils are a major source of dietary vitamin E (Goffman and Becker, 2002). Vitamin E occurs in the form of tocopherols, which are lipid-soluble antioxidants that accumulate in the chloroplast. Their function is to protect photosystem II from oxidative damage under the influence of free/released lipid peroxyl radicals and singlet oxygen (Quadrana et al., 2013) and in seeds they play role in preventing oxidation of polyunsaturated fatty acids (PUFAs). The four forms of tocopherol (α-, β-, γ- and δ-), vary in the number and position of methyl substituents attached to the chromanol ring (Munné-Bosh and Alegre, 2002). The most abundant forms of vitamin E in rapeseed oil are γ- and α-tocopherol, with a small proportion of δ-tocopherol (Fritsche et al., 2012, Wang et al., 2012). Besides its nutritional value, α-tocopherol is the most potent vitamin E, whereas the γ- and δ-tocopherol forms are valued for their oil-stabilizing properties (Munné-Bosh and Alegre, 2002), which is particularly relevant for PUFA-rich oils, such as rapeseed. Tocopherol content and composition in rapeseed varies widely; values for total tocopherol content (TTC) have been reported ranging between 166 and 687 mg.kg$^{-1}$, α-tocopherol content between 59 and 286 mg.kg$^{-1}$, and γ-tocopherol content from 107 to 280 mg.kg$^{-1}$. The ratio between α- and γ-tocopherol has also been reported to range between 0.33 and 2.14 (Dolde et al., 1999; Goffman and Becker, 2002; Wang et al., 2012; Fritsche et al., 2012). Genes involved in the tocopherol biosynthetic pathway have been identified in Arabidopsis thaliana and other model plants (Valentin et al., 2006; Endrigkeit et
al., 2009; Li et al., 2012), (Figure 1). QTL affecting seed tocopherol content and composition have also been reported (Gilliand et al., 2006), but the control of biosynthesis is poorly understood.

The first AT panel reported for B. napus (Harper et al., 2012) comprised only 84 accessions and was smaller than is usually required for association studies (Spencer et al., 2009), meaning that it could be used successfully only for traits with a simple genetic basis. In this study, we report the establishment of a full AT platform for the crop species B. napus, based on a widely-shared genetic diversity panel of 383 accessions, which can be used to address the genetic architecture of a broad range of traits. We validated the resource by analysing with the new platform a trait that had been analysed using the original panel (erucic acid content of seed oil), and a new trait (the relative content of γ and α forms of tocopherol in seeds).

Results

The RIPR genetic diversity panel

A diversity panel of 383 B. napus doubled haploid (DH) or inbred accessions was assembled, with the aim of covering the breadth of genetic variation available in the species. This panel included the breadth of crop types of B. napus and comprised 362 inbred lines previously used by Bus et al. (2011) and Harper et al. (2012) plus 21 further accessions as used by Thomas et al. (2016). The list of
accessions is shown in Data S1. The panel is named RIPR after the research project “BBSRC Renewable Industrial Products from Rapeseed (RIPR) Programme” that funded its development and genotyping.

**Functional genotypes**

Functional genotypes were produced for the panel based on leaf RNA, with 100-base read length mRNAseq data produced using the Illumina HiSeq 2000 platform. A total of $1.92 \times 10^{12}$ bases of sequence data were produced. The sequence reads were mapped to the CDS gene model-based *Brassica* AC pan transcriptome reference (He et al., 2015), which comprised 116,098 gene models, has an aggregate length of 118,657,829 bases and for which we provide an updated gene order based on a high density SNP linkage map as shown in Data S2. Sequence read mapping statistics are summarised in Data S1. Mean figures of 50,165,125 reads were generated per accession, with 32,275,718 being mapped across 61,620,266 bases of the reference sequence, representing 52.1-fold coverage of the 51.9% of the predicted transcriptome to which mRNAseq reads were mapped. SNPs were identified and gene expression quantified. Across the panel of 383 lines, 355,536 SNPs were scored, of which the majority (87.0%) were hemi-SNPs, as found in previous *B. napus* studies (Trick et al., 2009). A total of 127,153,561 allele calls were made, with 9,017,727 (6.6%) missing values. Significant expression (>0.4 RPKM) was detected for 53,889 CDS models (46.4% of all CDS models in the AC pan transcriptome reference), of which 25,834 belong to the A genome and 28,055 to the C genome.
The functional genotypes are available from the York Oilseed Rape Knowledgebase (http://www.yorknowledgebase.info).

**Genetic architecture of the Population**

The 355,536 SNP markers scored across the RIPR panel were used to analyse the relatedness of members of the panel. First, a distance matrix was generated and visualised by the dendrogram shown in Figure 2a. The assigned crop types (Data S1) show the expected clustering, as shown in Figure 2b. Next, the population structure of the panel was analysed using PSIKO (Popescu et al., 2014). The highest likelihood is a subpopulation K=2, with mixture across the panel as illustrated in Figure 2c. Finally, LD was calculated across the genome, as summarised in Figure S1, producing a mean value of 0.031 for the population.

**Seed erucic acid analysis**

Erucic acid is a 22-carbon monounsaturated fatty acid. Its content in rapeseed oil is one of the key determinants of suitability for use as an edible or industrial oil. Detection of the known loci controlling the biosynthesis of erucic acid in seeds was used as a validation study for the first report of AT (Harper et al., 2012). We re-analysed this trait to compare the performance of the original panel with the new RIPR panel. The fatty acid composition of seeds was determined for 376 lines of the RIPR diversity panel (summarised in Data S3). The erucic acid content of seeds varied between 0 and 51%, reflecting the range of crop types represented in the panel, which included modern Canola quality rapeseed varieties as well as crop types for which seed composition was not the subject of
active domesticated selection process (hence representative of “unimproved” seed composition).

**Associative Transcriptomics of erucic acid content**

The first stage of validation of the new AT platform for *B. napus* involved analysis of seed erucic acid content, a trait for which the two main control loci are known and were confirmed previously by AT (Harper et al., 2012). The estimated narrow-sense heritability ($h^2$) for the erucic acid trait was estimated from the SNP analysis as 0.794. 318 genome-assigned SNP markers above the Bonferroni-corrected significance threshold of $P = 0.05$ (i.e. $-\log_{10} P$ value of 6.7) were detected across association signals on chromosomes A5, A8, A9, A10 and C3 (Data S4, Figure S3), as illustrated in Figure 3a. The main loci controlling erucic content (on chromosomes A8 and C3) provide association signals with a significance eight orders of magnitude greater: $-\log_{10} P > 16$, compared with <8 in the previous study. The known control genes, orthologues of *FAE1* (AT4G34520), represented by gene models Cab035983.1 and Bo3g168810.1, are near the centres of these SNP association peaks, in the distance of 6 genes (~42 kb) and 9 genes (~56 kb) from the closest significantly associated gene, respectively, according of the reference sequence (Data S4). In addition, SNP associations were found for a region of the genome, on chromosome A5, which were not previously detected. This indicates the position of a novel locus with minor effect on the trait. A candidate for the trait control gene in this region is Cab033920.1. This gene is an orthologue of AT2G34770.1, which is annotated as fatty acid hydroxylase 1 which has a potential role in very long chain fatty acid biosynthesis. An association
signal was also detected for a relatively large region of chromosome A9, which we interpret as corresponding to a seed glucosinolate-controlling locus, which was co-selected in modern low erucic rapeseed cultivars to produce Canola quality seed.

In addition to association analysis using SNP markers, AT also reveals associations between gene expression markers (in the tissue of second true leaves used for the development of functional genotypes) and trait variation. In the case of seed erucic acid content, the main control genes (orthologues of \textit{FAE1}) are transcriptionally inactive in the tissue (leaves) sampled for production of the functional genotypes. However, we are still able to detect both SNP and GEM association peaks through markers in linkage disequilibrium (LD) with \textit{FAE1} on A8 and C3 as illustrated in Figure 3b. The lower resolution observed for the A8 peaks may reflect the influence of two strong bottlenecks during the breeding selection (Hasan et al., 2008) for low glucosinolate content (controlling loci on chromosome A2, A9, C2 and C9) and zero seed erucic acid (controlling loci on chromosome A8 and C3), or perhaps the presence of additional minor effect genes located on A8 that are also contributing to the erucic trait. Indeed there are many potential candidate genes in the region which could have an effect, including an orthologue of FAD6 (AT4G30950) which could act to reduce the pool of oleic acid available for elongation to erucic acid. In addition, there is a signature of slightly inflated LD on the first half of A8, which may further contribute to reducing the resolution of association peaks in this region (Figure S1).
The clear signals in the transcript abundance-based association analysis confirms the stability of differential gene expression across the panel and its utility for the identification of association signals. Regions of the genome previously associated with seed glucosinolate content (selected alongside erucic content in Canola quality rapeseed) show particularly strong transcript abundance associations, which we interpret as consequences of the extensive structural variation in these regions of the genome (He et al, 2016). The new AT platform generates strong signals due to the large, diverse panel and superior number of markers assigned to homoeologues, properties lacking in the platform reported previously (Harper et al, 2012).

Tocopherol phenotype analysis

We selected tocopherols in seeds as test traits of unknown genetic basis, quantifying α, γ and δ forms. Tocopherols were purified from seeds and quantified for 377 accessions of the RIPR panel. The results are summarised in Data S5 and Figure S2. Total tocopherol in seeds varied from 197 to 445 mg.kg\(^{-1}\), with the main types being γ-tocopherol (78 to 347 mg.kg\(^{-1}\)) and α-tocopherol (51 to 229 mg.kg\(^{-1}\)), the relative proportions of which (measured as the γ/α-tocopherol ratio) varied greatly, ranging from 0.485 to 5.00, δ-tocopherol was a minor component (1.8 to 9.9 mg.kg\(^{-1}\)). Analysis of tocopherol characteristics by crop type showed that γ-tocopherol content tended to be higher in spring crop types and α-tocopherol content tended to be higher in winter crop types, as illustrated in Figure 2d.
Given that the purpose of tocopherols in seed oil is to protect against oxidation, we assessed the diversity panel for correlation of tocopherol traits with the proportions of the fatty acids found in seed oil that are most susceptible to oxidation, the polyunsaturated fatty acids (PUFAs) linoleic and linolenic. The content of these fatty acids had been determined alongside that of erucic acid (Data S3). A weak positive correlation between total tocopherol and PUFA content was, indeed, identified ($R^2 = 0.13; p<0.001$).

**Associative Transcriptomics of tocopherol composition**

To undertake AT for tocopherol traits, we analysed the population for loci controlling the proportion of tocopherol occurring in the $\gamma$ form rather than the $\alpha$ form by using $\gamma/\alpha$ ratio as the trait. The SNP-based association analysis, as illustrated in Figure 4a, revealed exceptionally strong associations with markers in a very small regions of chromosome C2, along with weaker associations with a few markers in regions of chromosomes A2 and A10. Unlike seed erucic acid, tocopherol composition has not been selected by *B. napus* breeders. We interpret the very sharp association signal as indicative of this lack of selection and to be consistent with LD across most of the genome. The association peak on chromosome C2 includes 33 genome-assigned markers above the Bonferroni-corrected significance threshold ($\alpha = 0.05; – \log_{10}P$ value of 6.7) (Data S6, Figure S3). These delineated a genomic region containing 39 genes, including an orthologue of *VTE4*, which encodes a $\gamma$-tocopherol methyl transferase ($\gamma$-TMT), an enzyme that converts $\gamma$-tocopherol into $\alpha$-tocopherol (Figure 1). A homoeologous region including a duplicate copy of *VTE4* gene within association
peak on chromosome A2 was observed, while there was no obvious candidate
gene in the region of chromosome A10 showing associations. Four transcript
abundance-based markers above the Bonferroni-corrected significance
threshold (− log\text{10} P value of 6.03 for GEMs) were identified on chromosome C2,
C5 and C7 (Figure 4b). The identification of the gene VTE4 as the most highly
associated GEM on chromosome C2 demonstrated the ability for AT to efficiently
provide candidate genes associated with traits of interest.

To investigate whether the top selected markers are predictive for γ/α ratio, we
performed a set of “take-one-out” permutations for the SNP and GEM markers
identified from association analysis of 377 accessions adapted from Harper et al
(2016). Markers above the Bonferroni line (Data S6 and S7) were selected for
each round of permutations. For SNP data, the allelic effects of each of these
markers was used to predict trait values for the missing accessions based on
their scored genotypes. For GEM data, RPKM values were fitted to the regression
line to predict trait values. The predicted trait values against the observed trait
are illustrated, as scatter plots, in Figure 5 and confirmed excellent predictive
ability (R^2 = 0.59 for SNPs and R^2 = 0.47 for GEMs between predicted and
observed values; p<0.001), which reflect the estimated narrow-sense heritability
(h^2) of 0.452 for γ/α ratio. These SNPs and GEMs can therefore be used as
promising markers in marker assisted breeding.
In order to confirm the role of the VTE4 orthologue in the associated region of C2 (Bo2g050970.1), we used the transcript quantification data that were obtained alongside the transcriptome SNP data as part of the functional genotypes. As illustrated in Figure 6, these show that the expression level of Bo2g050970.1 in the tissue sampled to produce the functional genotypes (leaves) is negatively correlated with the γ/α ratio ($R^2 = 0.41$, $p<0.001$). This is consistent with the predicted γ-TMT activity of the gene encoded by Bo2g050970.1 (i.e. lower expression leading to less conversion of γ-tocopherol to α-tocopherol). There had been no significant associations between SNPs within Bo2g050970.1 and the γ/α ratio, consistent with the basis of the allelic variation being variation in gene expression rather than variation in gene sequence.

**Discussion**

Association studies are becoming increasingly widely-used in crops for identifying molecular markers linked to trait-controlling loci (Rafalski, 2010). However, polyploid crops present additional difficulties that must be overcome, including the intrinsic genome complexity and increased genome structural instability, such as the copy-number variations (CNV) which affect gene families (Zhang et al., 2013; Renny-Byfield and Wendel, 2014). Such difficulties occur in *B. napus*, as was recently shown by Chalhoub et al. (2014) and He et al. (2016). Association studies have to meet many demands to maximize the probability of identifying marker-trait associations. In addition to good planning of experimental design,
along with access to all the necessary equipment and available funds, there is also the need to choose a permanent and sufficiently large set of diverse and preferably homozygous individuals, the larger size and higher genetic diversity of which providing sufficient power for association analysis (Spencer et al., 2009; Huang and Han, 2014). Once assembled, association panels need to be genotyped with molecular markers in a sufficiently high density to identify polymorphisms in linkage disequilibrium with trait-controlling loci. The development of suitable association panels is challenging for individual research groups, providing a driver for the development of community resources.

In this study, we introduce a new genetically diverse AT panel of 383 rapeseed accessions, together with a mapping platform that comprises complete genotype information for this panel, which may be used for a broad range of association studies suitable for re-phenotyping any trait, without the need of additional genotyping. This panel, being made available with all transcriptomic data, offers a large range of potential applications: identifying causative genes, uncovering unknown pathways, identifying regulatory genes or transcription factors, and screening of available germplasm for allelic variants and to support the development of molecular markers for marker-assisted breeding. Our resource provides 355,536 SNP markers, equivalent to one SNP every 0.33 kb across our *Brassica napus* AC pan-transcriptome reference. The SNP density is much higher than the density of the commercially available 60K Brassica Infinium® SNP array, which only provided 26,841 or 21,117 SNPs for recent *B. napus* GWAS studies (Li et al., 2014, Xu et al., 2016). Although the number of SNPs
can even be greater when using whole genome re-sequencing, as shown by Huang et al. (2013), the advantage of transcriptome re-sequencing using mRNAseq is the availability of transcript abundance data; in our case for 46% of the genes present in the AC pan-transcriptome reference sequence. In this study, we demonstrate a significant step-change in resolution from our original AT platform based on a panel of 84 accessions, as reported in Harper et al. (2012). The unigene-based transcriptome reference sequence used by that platform had relatively poor capability to resolve homoeologous loci, due to its construction based on a Brassica-wide transcriptome assembly and subsequent “curing” to more closely match the progenitor genomes. In the absence of the ability to map sequence reads unambiguously to the correct homoeologue, most SNPs appear, due to cross-mapping, as “hemi-SNPs”, i.e. where one allele comprises a mixture of two bases (Trick et al., 2009). In the original platform only a small proportion of markers could be assigned with high confidence to a genome, the majority being assigned to both homoeologous positions. The new platform is based mainly on gene models originating from the genome sequences of the progenitor species permits more discriminating read mapping, resulting in a greater proportion of “simple SNPs” (i.e. where the polymorphism is between resolved single bases only), which can be assigned with confidence to a genome. Where there are association peaks comprising pale points in homoeologous positions to the associations identified, such as those observed in regions of A2 depicted in Figure 4a, these can be disregarded as homoeologous “shadows” of the regions genuinely containing causative variation. SNP discovery for particular genes from juvenile leaves can be limited by their transcription in different phenological stage
or tissue, but candidate locigenes associated with trait manifesting in different
time or place can be still identified, as demonstrated here in case of FAE1 and in
previous AT studies (Lu et al., 2014; Wood et al., 2017). This is possible due to
the presence of variation in genes in LD with the causative gene, resulting in an
associated region including the control gene. In addition, the new platform
provides much greater resolution of the contributions to the transcriptome of pairs
of homoeologous genes. This permitted efficient detection of association peaks
based solely on transcript abundance variation, as illustrated in Figure 3.
Moreover, the current platform also allows deeper insight of the structural
changes and functional interactions between B. napus AC genomes. Information
about respective homologous genes including their copy number, sequence
variation and transcript prevalence provides important information in polyploid
research.

In addition to extending previous association studies of the control of seed erucic
acid content, a trait selected recently by rapeseed breeders, we applied the
platform to a trait not previously selected by breeders, or studied extensively: the
control of tocopherol (Vitamin E) forms accumulated in seeds. We analysed seed
tocopherols in 377 rapeseed accessions for their type and content. The profiles
presented here showed a high degree of variability for the γ-/α-tocopherol ratio
(CV=53%), displaying distinct patterns for different crop types, which allowed us
to identify gene Bo2g050970.1 (an orthologue of the Arabidopsis gene VTE4) on
chromosome C2 as a candidate gene, based on inference of gene function based
on studies of its orthologue in A. thaliana. Although there was no evidence of the
presence any specific allelic form of the VTE4 orthologue associated with \(\gamma/\alpha\)-tocopherol ratio, this gene has been easily identifiable by the presence of SNPs in surrounding genes. This set of tightly linked markers exhibited excellent predictive ability (Figure 5), which we attribute to the broad (species-wide) range of genetic variation represented by the RIPR diversity panel, overcoming the lack of predictive capability that can be encountered when applying markers to test material (Bush and Moore, 2012). The association we observed between transcript abundance of Bo2g050970.1 in leaves and the tocopherol \(\gamma/\alpha\) ratio in seed is consistent with our understanding that tocopherols are synthesized and localized in plastids and accumulate in all tissues with generally highest content in seeds (Sattler et al., 2004). In Arabidopsis, \(\gamma\)-TMT (VTE4, AT1G64970) is known to use \(\delta\)- and \(\gamma\)-tocopherols as substrates to produce \(\beta\)- and \(\alpha\)-tocopherols respectively (Shitani and DellaPenna, 1998) and the effect of VTE4 gene from B. napus on \(\alpha\)-tocopherol content has been also proved by overexpression in soybean and Arabidopsis (Chen et al., 2012, Endrigkeit et al., 2009).

By assembling and developing functional genotypes (i.e. comprising both gene sequence variation and gene expression variation) for a diversity panel representing species-wide genetic diversity, we have established a resource for the whole rapeseed research community to use. Furthermore, the success of the approach of Associative Transcriptomics for the identification not only of linked markers, but of candidates for causative genes, serves as an exemplar for plant and crop science more broadly.
Experimental procedures

Growth of the genetic diversity panel

The panel of 383 B. napus accessions is available from the John Innes Centre, Norwich, UK. It was planted in a randomized block design of five biological replicates under controlled conditions of two polytunnels at University of Nottingham as described by Thomas et al. (2016). The accessions comprise inbred derivatives of both recent and historic varieties and some research lines. Plants were bagged before flowering to prevent cross-pollination. Seeds were collected from individual plants at maturity. Seeds from 377 and 376 accessions were used for the tocopherol and erucic acid measurement, respectively. Based on descriptors originally received with the material and analysis of relatedness, they were attributed to one of seven different groups, namely spring oilseed rape (123), semi-winter oilseed rape (11), swede (27), kale (3), fodder (6), winter oilseed rape (169) or crop type not assigned (44), as listed in Data S1.

Measurement of fatty acid content and composition

For fatty acid methyl esters (FAME) analysis, 30 mg of seeds were homogenized in a glass vial with 5 ml of heptane. To the homogenate, 500 µl of 2 M potassium hydroxide was added, left for one hour and neutralised with sodium hydrogen sulphate monohydrate. The upper phase was transferred into a crimp cap Chromacol 0.8 ml vials for analysis using a DANI Master GC fitted with an SGE-BPX70 double column.
Measurement of tocopherol content and composition

The α-, γ- and δ-tocopherol (the sum of which formed total tocopherol, TTC) were extracted from a homogenous mixture of 80 mg rapeseed seeds and analyzed by normal-phase HPLC as described previously (Fritsche et al., 2012). Modified mobile phase A was heptane (Rathburn, Walkerburn, UK), phase B heptane:dioxane (Sigma-Aldrich, Gillingham, UK) (90:10, v/v). Internal standard, α-tocopherol acetate (Sigma-Aldrich), was added to each sample at a concentration of 25.4 µM (12 µg·mL⁻¹).

SNP identification and Transcript quantification for RNA-seq data

The growth conditions, sampling of plant material, RNA extraction and transcriptome sequencing was carried out as described by He et al. (2016). The RNA-seq data from each accession line were mapped on to recently-developed ordered Brassica A and C pan-transcriptomes (He et al., 2015) as reference sequences Maq v0.7.1 (Li et al., 2008). SNPs were called by the meta-analysis of alignments as described in Bancroft et al. (2011) of mRNAseq reads obtained from each of the B. napus accessions. SNP positions were excluded if they did not have a read depth in excess of 10, a base call quality above Q20, missing data below 0.25, and 3 alleles or fewer. An additional noise threshold was employed to reduce the effect of sequencing errors, whereby ambiguous bases were only allowed to be called if both bases were present at a frequency of 0.2 or above. This resulted in a set of 355,536 SNPs, of which 256,397 had the second most frequent allele in the population, so called here as a minor allele.
frequency (MAF) > 0.01. The markers were also classified as those that can be
assigned with confidence to the genomic position of the CDS model in which they
are scored (simple SNPs and hemi-SNPs genetically mapped into the
appropriate genome using the TNDH mapping population), and those that cannot
as the polymorphism may be in either homoeologue of the CDS model in which
they are scored (hemi-SNPs not genetically mapped into the appropriate genome
using the TNDH mapping population). Transcript abundance was quantified and
normalized as reads per kb per million aligned reads (RPKM) for each sample for
116,098 CDS models of the pan-transcriptome reference. Significant expression
(>0.4 RPKM) was detected for 53,889 CDS models.

**Clustering based on SNP genotypes**

Clustering and dendrogram visualisation on SNP data was performed by in-
house R script. R package “phangorn” was used for generating distance matrix
with JC69 model (Schliep, 2011).

**Assessment of LD**

Pairwise linkage disequilibrium was calculated and heatmaps produced for each
individual chromosome, and these values used to calculate the mean LD across
the genome. SNPs were removed from the analysis if they were not confirmed
by TNDH population (Qiu et al., 2006) that assigned to the A or C genome and if
their minor allele frequency was below 0.01. A single SNP was selected at
random from each CDS model to reduce the effect of many linked SNPs in the
same gene. Pairwise r2 LD matrices and heatmaps were calculated for each chromosome using the R package LDheatmap (version 0.99-2; Shin et al., 2006).

**Associative Transcriptomic analysis**

SNPs and gene expression markers (GEMs) association analysis was performed using R as previously described by (Harper et al., 2012, Sollars et al., 2017), with modifications: to deal with the greatly increased sizes of the datasets, PSIKO (Popescu et al., 2014) was used for Q-matrix generation and GAPIT R package with a mixed linear model (Lipka et al., 2012) was used for GWAS analysis. For SNP association Manhattan plots, SNP markers were filtered to include only those with minor allele frequency > 0.01, markers that could be assigned with confidence to the genomic position of the CDS model are rendered as dark points and markers that could not be assigned with confidence were rendered as pale points. For GEM association, CDS models were filtered prior to regression to include only those with mean expression across the panel > 0.4 RPKM. The association between gene expression and traits was calculated by fixed effect linear model in R with RPKM values and the Q matrix inferred by PSIKO as the explanatory variables and trait score the response variable. R² regression coefficients, constants and significance values were outputted for each regression. Genomic control (Devlin and Roeder, 1999) was applied to the GEM analysis to correct for spurious associations, with p-value adjustment applied when the genomic inflation factor (λ) was observed to be greater than 1.

**Validation of marker association by trait prediction**
The predictive power of the best GEMs and SNPs were assessed using a “take one out” approach (Harper et al. 2016) whereby each accession is removed from the SNP or GEM analysis in turn. An in-house R script was performed with adaptation from Harper 2016, with a modification of incorporating all SNPs and GEMs above bonferroni lines. When permutations finishes, an r square value is calculated from predicted trait values regressed against the observed trait values which indicates the predictive power of the top selected GEMs and SNPs.

**Accession numbers**

Sequence data from this article can be found in the SRA data library under accession number PRJNA309367.

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**Supporting Information**
Supporting data are provided. The largest datasets, representing the functional
genotypes of the RIPR panel, are accessible via a data distribution website:
http://www.yorkknowledgebase.info/. The smaller datasets accompany the
manuscript, as MS Excel files:

Supporting figures:
Figure S1. Genome-wide Linkage Disequilibrium analysis for the RIPR diversity
panel: Figure S1_LD_SAF_1perc_26-9-17.pdf

Figure S2. Histograms of seed tocopherol composition of the RIPR diversity
panel in different crop types: Figure S2_histograms of seed tocopherol
composition.pdf

Figure S3. QQ plots from GEM and SNP association analysis for erucic acid
and ω/α tocopherol ratio: Figure S3_QQ_plots.pdf

Supporting data:
Data S1. List of cultivars, crop type classifications and Illumina read mapping
statistics: Data S1_cultivars and read mapping_20-12-16.xlsx.

Data S2. Ordered list of CDS gene model-based Brassica AC pan
transcriptome: Data S2_v11 pan-tanscriptome_20-12-16.xlsx.
Data S3. Seed fatty acid composition of the RIPR diversity panel: Data S3_fatty acids_10-04-17.xlsx.

Data S4. Markers and genomic regions showing association with variation for erucic acid content: Data S4_erucic-associated regions_30-3-17.xlsx.

Data S5. Seed tocopherol composition of the RIPR diversity panel: Data S5_tocopherols_14-10-16.xlsx.

Data S6. Markers and genomic regions showing association with variation for γ/α tocopherol ratio: Data S6_tocopherol-associated regions_SNPs.xlsx.

Data S7. Gene expression markers showing association with variation for γ/α tocopherol ratio: Data S7_tocopherol-associated regions_GEMs.xlsx

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**Figure legends**

Figure 1. Simplified tocopherol biosynthesis pathway in plants. HPP, p-hydroxyphenylpyruvate; HGA, homogentisic acid; MPBQ, 2-methyl-6-phytyl-1,4-benzoquinone; DMPBQ, 2,3-dimethyl-5-phytyl-1,4-benzoquinone; PDP, phytol-diphosphate; HPPD, HPP dioxygenase; VTE1, tocopherol cyclase; VTE2, homogentisate phytyltransferase; VTE3, MPBQ methyltransferase; VTE4, γ-tocopherol methyltransferase; VTE5, phytol kinase.

Figure 2. Population structure and trait variation across the RIPR panel. A. Relatedness of accessions in the panel based on 355,536 scored SNPs. B. Main crop types in the panel, colour-coded: orange for spring oilseed rape, green for semi-winter oilseed rape, light blue for swede, dark blue for kale, black for fodder and red for winter oilseed rape, grey for crop type not assigned. C. Population structure for highest likelihood K = 2. D. Variation for seed content of α-tocopherol (light blue), γ-tocopherol (dark blue) and δ-tocopherol (magenta).
Figure 3. Association analysis. A. Transcriptome SNP markers with seed erucic acid content. The SNP markers are positioned on the x-axis based in the genomic order of the gene models in which the polymorphism was scored, with the significance of the trait association, as –log10P, on the y-axis. A1 to A10 and C1 to C9 are the chromosomes of *B. napus*, shown in alternating black and red colours to permit boundaries to be distinguished. Hemi-SNP markers (i.e. polymorphisms involving multiple bases called at the SNP position in one allele of the polymorphism) for which the genome of the polymorphism cannot be assigned are shown as light points whereas simple SNP markers (i.e. polymorphisms between resolved bases) and hemi-SNPs that have been directly linkage mapped, both of which can be assigned to a genome, are shown as dark points. The broken light blue horizontal line marks the Bonferroni-corrected significance threshold of 0.05.

B. Transcript abundance with seed erucic acid content. The gene models are positioned on the x-axis based in their genomic order, with the significance of the trait association, as –log10P, on the y-axis. The broken dark blue horizontal line marks the 5% false discovery rate.

Figure 4. Association analysis. A. Transcriptome SNP association analysis for seed γ/α tocopherol ratio. The SNP markers are positioned on the x-axis based in the genomic order of the gene models in which the polymorphism was scored, with the significance of the trait association, as –log10P, on the y-axis. A1 to A10 and C1 to C9 are the chromosomes of *B. napus*, shown in alternating black and
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B. Association analysis of transcript abundance with seed γ/α tocopherol ratio.

The gene models are positioned on the x-axis based in their genomic order, with the significance of the trait association, as $-\log_{10} P$, on the y-axis. The broken dark blue horizontal line marks the 5% false discovery rate.

Figure 5. Test of the predictive ability of SNP and GEM markers associated with γ/α tocopherol ratio by using "take-one-out" permutation. The allelic effects of each of 36 SNP markers associated with γ/α tocopherol ratio was used to predict the γ/α tocopherol ratio for the missing accessions. For GEM data, RPKM values for each of 4 GEMs were fitted to the regression line to predict γ/α tocopherol ratio. The strong correlation between predicted and observed γ/α tocopherol ratio values ($R^2 = 0.59; p<0.001$ for SNPs and $R^2 = 0.47; p<0.001$ for GEMs) demonstrates excellent predictive ability.

Figure 6. Relationship between expression in leaves of Bo2g050970.1 and the tocopherol γ/α ratio in seed. The ratio of γ-tocopherol / α-tocopherol measured in
seeds was regressed against the transcript abundance in leaves of the VTE4 orthologue Bo2g050970.1 (R²=0.26; p<0.001), measured as reads per kilobase per million aligned reads (RPKM).
Figure 1. Simplified tocopherol biosynthesis pathway in plants. HPP, p-hydroxyphenylpyruvate; HGA, homogentisic acid; MPBQ, 2-methyl-6-phytyl-1,4-benzoquinone; DMPBQ, 2,3-dimethyl-5-phytyl-1,4-benzoquinone; PDP, phytyl-diphosphate; HPPD, HPP dioxygenase; VTE1, tocopherol cyclase; VTE2, homogentisate phytyltransferase; VTE3, MPBQ methyltransferase; VTE4, γ-tocopherol methyltransferase; VTE5, phytol kinase.
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B. Association analysis of transcript abundance with seed γ/α tocopherol ratio. The gene models are positioned on the x-axis based in their genomic order, with the significance of the trait association, as $-\log_{10}P$, on the y-axis. The broken dark blue horizontal line marks the 5% false discovery rate.
Figure 5. Test of the predictive ability of SNP and GEM markers associated with \( \gamma/\alpha \) tocopherol ratio by using "take-one-out" permutation. The allelic effects of each of 36 SNP markers associated with \( \gamma/\alpha \) tocopherol ratio was used to predict the \( \gamma/\alpha \) tocopherol ratio for the missing accessions. For GEM data, RPKM values for each of 4 GEMs were fitted to the regression line to predict \( \gamma/\alpha \) tocopherol ratio. The strong correlation between predicted and observed \( \gamma/\alpha \) tocopherol ratio values \( (R^2 = 0.59; p<0.001 \text{ for SNPs and } R^2 = 0.47; p<0.001 \text{ for GEMs}) \) demonstrates excellent predictive ability.
Figure 6. Relationship between expression in leaves of Bo2g050970.1 and the tocopherol γ/α ratio in seed. The ratio of γ-tocopherol / α-tocopherol measured in seeds was regressed against the transcript abundance in leaves of the VTE4 orthologue Bo2g050970.1 (R²=0.41; p<0.001), measured as reads per kilobase per million aligned reads (RPKM).