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Genetical and Comparative Genomics of *Brassica* under Altered Ca Supply Identifies *Arabidopsis* Ca-Transporter Orthologs

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Although Ca transport in plants is highly complex, the overexpression of vacuolar Ca2+ transporters in crops is a promising new technology to improve dietary Ca supplies through biofortification. Here, we sought to identify novel targets for increasing plant Ca accumulation using genetical and comparative genomics. Expression quantitative trait locus (eQTL) mapping to 1895 cis- and 8015 trans-loci were identified in shoots of an inbred mapping population of *Brassica rapa* (IMB211 × R500); 23 cis- and 948 trans-eQTLs responded specifically to altered Ca supply. eQTLs were screened for functional significance using a large database of shoot Ca concentration phenotypes of *Arabidopsis thaliana*. From 31 *Arabidopsis* gene identifiers tagged to robust shoot Ca concentration phenotypes, 21 mapped to 27 *B. rapa* eQTLs, including orthologs of the Ca2+ transporters At-CAX1 and At-ACA8. Two of three independent missense mutants of *BraA.CAX1a*, isolated previously by targeting induced local lesions in genomes, have allele-specific shoot Ca concentration phenotypes compared with their segregating wild types. *BraA.CAX1a* is a promising target for altering the Ca composition of *Brassica*, consistent with prior knowledge from *Arabidopsis*. We conclude that multiple-environment eQTL analysis of complex crop genomes combined with comparative genomics is a powerful technique for novel gene identification/prioritization.

INTRODUCTION

Ca is an essential element required in large amounts by plants for structural roles and also, in lesser amounts, for cell signaling (White and Broadley, 2003; Dodd et al., 2010; White, 2014). Leaf/shoot Ca concentration (hereafter referred to as shoot-Ca) is typically in the low percentage (w/w) dry weight range, although it varies widely among terrestrial species (0.02 to 13.1%; White and Broadley, 2003; Dodd et al., 2010; Martinoia et al., 2012; White, 2014). At the organ scale, Ca accumulates preferentially within actively transpiring organs because it is immobile in the phloem (White and Broadley, 2003; Karley and White, 2009). The initial entry of Ca to plants through roots and the transfer to xylem tissues...
involve a combination of symplastic, transcellular, and apoplastic routes (White, 2001, 2014; White and Broadley, 2003), which are affected by apoplastic barriers currently being characterized (Baxter et al., 2009; Hosmani et al., 2013).

Shoot-Ca is under strong environmental and genetic control. Unsurprisingly, plants typically have greater shoot-Ca at high external Ca supply \([\text{Ca}_{\text{ext}}]\) than at low \([\text{Ca}_{\text{ext}}]\) (White, 2014). In Brassicaceae species, Ca sequestration in mesophyll vacuoles occurs at high \([\text{Ca}_{\text{ext}}]\) (Conn and Gillilah, 2010; Conn et al., 2011, 2012; Rios et al., 2012). However, relationships between shoot-Ca and \([\text{Ca}_{\text{ext}}]\) differ between “physiotypes” (Kinzel, 1982) associated with certain plant families. Thus, shoot-Ca responds to increased \([\text{Ca}_{\text{ext}}]\) to a greater extent in calcicole-like calcitrophic plants (e.g., Brassicaceae and Crassulaceae) than among calcifuge-like potassium plants (e.g., Apiaceae and Asteraceae) and oxalate plants (e.g., Amaranthaceae (Kinzel, 1982; Kinzel and Lechner, 1992; White and Broadley, 2003; White, 2014). Genetic controls of shoot-Ca have been quantified more formally at contrasting scales. For example, at higher levels of phylogenetic organization, shoot-Ca is low among commelinoid monocot species (e.g., grasses and sedges), as is shoot Mg concentration (shoot-Mg). Species of Amaranthaceae (e.g., spinach [Spinacia oleracea]) and related clades of the order Caryophyllales have low shoot-Ca but high shoot-Mg (Broadley et al., 2003, 2004; Watanabe et al., 2007; Broadley and White 2012). Phylogenetic variation in shoot-Ca may be due in part to evolutionary differences in cell wall chemistry, as up to 50% of Ca in plants is bound to pectins in cell wall lamellae (White and Broadley, 2003). At intraspecific scales, shoot-Ca is strongly heritable among leafy Brassica oleracea subspecies, with several significant quantitative trait loci (QTLs) being identified under controlled environment and field conditions (Broadley et al., 2008). In addition to natural genetic variation, functional studies have revealed key roles for specific gene products in regulating the vacuolar sequestration of Ca in the leaf mesophyll (Cheng et al., 2005; Conn et al., 2011) and the root uptake of Ca (Baxter et al., 2009) and their effects on shoot-Ca. For example, Arabidopsis thaliana \(cax1\) \(cax3\) double mutants have lower shoot-Ca (Cheng et al., 2005; Conn et al., 2011) due to reduced uptake into mesophyll vacuoles. Conversely, ectopic expression of a truncated \(At-CAX1\) (\(sCAX1\)) increased shoot-Ca in carrot (Daucus carota), potato (Solanum tuberosum), tomato (Solanum lycopersicum), and tobacco (Nicotiana tabacum; Shigaki and Hirschi, 2006; Morris et al., 2008).

Despite advances in our knowledge of environmental and genetic factors controlling shoot-Ca at multiple scales, there are no reported comprehensive studies of these factors and their interactions (genetic × environment). Recent developments in genetical genomics, also known as expression quantitative trait locus (eQTL) analysis, can now measure variation in global gene expression among individuals within mapping populations (Kliebenstein, 2009; Duka et al., 2010; Hammond et al., 2011; Holloway et al., 2011; Moscou et al., 2011; Cubillos et al., 2012; Ballini et al., 2013; Huang et al., 2013). Variation in gene expression can be linked to sequence polymorphisms in target genes or, perhaps more interestingly, inherited differences in cis-regulatory (proximal) or trans-regulatory (distal) regions. eQTLs, therefore, can be used to integrate physical and genetic maps, to select genes underlying phenotypic differences for detailed characterization, and/or to identify novel regulatory networks. However, despite the increasing affordability of genetical genomic approaches, reported multiple-environment eQTLs are still limited to a few studies, such as two-condition biotic (Duka et al., 2008; Chen et al., 2010; Moscou et al., 2011) and abiotic (Hammond et al., 2011) experiments.

The primary aim of this study was to characterize the genetic, environmental, and genetic × environment effects on plant gene regulation by identifying eQTLs under altered \([\text{Ca}_{\text{ext}}]\) supply in Brassica rapa \((2n = 2x = 20)\) and characterizing candidate genes. This species can accumulate relatively high Ca concentrations in its leaf mesophyll tissues (Rios et al., 2012). Furthermore, B. rapa has a sequenced genome that enables comparative genomic analyses with Arabidopsis (Wang et al., 2011) and postgenomic resources, including high-density exon-based expression arrays (Love et al., 2010), populations of mutants identified by targeting induced local lesions in genomes (TILLING) (Stephenson et al., 2010; Wang et al., 2012), and mapping lines (Iniguez-Luy et al., 2009). All of these resources were exploited in this study to test the potential of a Brassica-to-Arabidopsis-to-Brassica workflow. Multiple-environment conditions were used so that cis- and trans-eQTLs, which were constitutive or responded solely to Ca or Mg or to both elements, could be identified. Novel eQTLs were characterized in silico from phenotypes of loss-of-function Arabidopsis mutants in a large public ionomics database (Baxter et al., 2007). Mutations in the Ca\(^{2+}\) transporter \(BraA.CAX1a\), identified previously by TILLING (Ó Lochlainn et al., 2011) and characterized here, show a link between eQTLs and altered shoot-Ca in B. rapa and Arabidopsis. Given that Brassica species are consumed widely as vegetable crops and are possible target crops for increasing intakes of important nutrients such as Ca (Broadley and White, 2010), this study reveals further novel candidate genes that might also be used for biofortification.

**RESULTS AND DISCUSSION**

**Responses of B. rapa to Altered \([\text{Ca}_{\text{ext}}]\) and External Mg Supply**

We investigated the response of shoot biomass and shoot-Ca to altered \([\text{Ca}_{\text{ext}}]\) and external Mg supply \([\text{Mg}_{\text{ext}}]\) (Figures 1 and 2). These responses were determined using an experimental design that included Ca and Mg supply as treatment factors, each represented by six external concentrations (i.e., plants were grown under 36 \([\text{Ca}_{\text{ext}}] \times [\text{Mg}_{\text{ext}}]\) conditions). Mg was used as a treatment factor so that responses specific to \([\text{Ca}_{\text{ext}}]\) could be separated from those due to elevated concentrations of a nutrient cation or chloride. There was a significant positive relationship between added and measured (water-extractable) compost \([\text{Ca}_{\text{ext}}]\) and \([\text{Mg}_{\text{ext}}]\) indicating that appropriate experimental conditions were used (Supplemental Figure 1). All primary biomass and tissue concentration data are presented in Supplemental Data Set 1.

There were no significant interactions of \([\text{Ca}_{\text{ext}}] \times [\text{Mg}_{\text{ext}}]\) on shoot biomass production (\(P > 0.05\)); hence, the data in Figure 1 are presented as global means for each main factor of \([\text{Ca}_{\text{ext}}]\) and \([\text{Mg}_{\text{ext}}]\). Shoot fresh and dry weights both increased when \([\text{Ca}_{\text{ext}}]\) increased from 0 to 9 mM (Figures 1A and 1C). The percentage dry weight (w/w) of shoots decreased when \([\text{Ca}_{\text{ext}}]\) increased from 0 to 9 mM and did not change significantly with further
increased by $\text{Mg}_{\text{ext}}$ and decreased with increasing Mg supply, although there was no significant effect on biomass or percentage dry weight in the range 0 to 9 mM $\text{Mg}_{\text{ext}}$ (Figures 1B, 1D, and 1F).

There was a general increase in shoot-Ca with increasing $\text{Ca}_{\text{ext}}$ across all $\text{Mg}_{\text{ext}}$ treatments (Figure 2). In IMB211 (Figures 2A, 2D, and 2G) and R-o-18 (Figures 2C, 2F, and 2I), shoot-Ca increased by >10-fold in response to increasing $\text{Ca}_{\text{ext}}$ and shoot-Ca exceeded 2% (w/w) at 30 mM $\text{Ca}_{\text{ext}}$ at low $\text{Mg}_{\text{ext}}$. In R500 (Figure 2B), shoot-Ca increased to ~1.5% (w/w) at 30 mM $\text{Ca}_{\text{ext}}$. At high $\text{Ca}_{\text{ext}}$ (15 and 30 mM), a high $\text{Mg}_{\text{ext}}$ led to a substantial decrease in shoot-Ca in IMB211 and R-o-18 but not in R500 (Figure 2). There was a general increase in shoot-Mg with increasing $\text{Mg}_{\text{ext}}$ across all $\text{Ca}_{\text{ext}}$ treatments (Supplemental Figure 2). Shoot-Mg increased by ~5-fold in response to increasing $\text{Mg}_{\text{ext}}$, with shoot-Mg reaching 1% (w/w) at 30 mM $\text{Ca}_{\text{ext}}$ at low $\text{Mg}_{\text{ext}}$ (Supplemental Figures 2D to 2F). At higher $\text{Mg}_{\text{ext}}$ (30 mM), increasing $\text{Ca}_{\text{ext}}$ led to a substantial decrease in shoot-Mg in IMB211 and R-o-18 but not in R500 (Supplemental Figures 2D to 2F). A 2 × 2 combination of $\text{Ca}_{\text{ext}} \times \text{Mg}_{\text{ext}}$ treatments was selected for subsequent eQTL experiments, defined as low/low (LL), high/low (HL), low/high (LH), and high/high (HH). The high concentrations were 24 mM $\text{Ca}_{\text{ext}}$ and 15 mM $\text{Mg}_{\text{ext}}$, and the low concentrations were 3 mM $\text{Ca}_{\text{ext}}$ and 1 mM $\text{Mg}_{\text{ext}}$.

**Figure 1.** Pooled Yield Responses of *B. rapa* as Functions of $\text{Ca}_{\text{ext}}$ and $\text{Mg}_{\text{ext}}$.

Plants were grown in compost for 18 d. Data are means of 54 samples: 3 genotypes (IMB211, R500, and R-o-18), 6 $\text{Mg}_{\text{ext}}$ or $\text{Ca}_{\text{ext}}$ treatment levels per $\text{Ca}_{\text{ext}}$ or $\text{Mg}_{\text{ext}}$ treatment level, respectively, and 3 replicates. LSD values (P = 0.05) are as follows: 0.1368 (shoot fresh weight [FW]; [A] and [B]), 0.0055 (shoot dry weight [DW]; [C] and [D]), and 2.634 (shoot percentage dry weight [%DW]; [E] and [F]).

**Genetical Genomics Analysis of Plant Responses to Ca and Mg**

**Novel Gene Expression Markers and Genetic Map for *B. rapa***

Sequence polymorphisms between the parents of the mapping population result in differences in the hybridization of transcripts to the probes on the array and subsequent differences in signal values for those probes. These differences can be used to develop markers based on gene expression. Initially, the expression data from the two parents, across all treatments, were analyzed to identify exon-level probe sets that had significantly different signal values. The linkage map reported here, and used for eQTL analyses, is based on 134 gene expression markers (GEMs), selected from 3272 potential GEMs, and mapped for 85 individuals from the *Brassica* *rapa* IMB211 × R500 Recombinant Inbred (BraiRRI) mapping population (Iniguez-Luy et al., 2009). This new map has a length of 811.6 centimorgan (cM), similar to that reported previously by Hammond et al. (2011), which spanned 881.6 cM. Of the 134 GEMs in the current map, 121 have reported genomic locations consistent with their mapping positions. An additional two have locations near the genetic position reported, but not in the same order as expected from the physical location. The remaining nine GEMs could not be located uniquely in the reported physical map at present. The median heritability of GEMs was 0.86, consistent with the high heritability of GEMs identified within this mapping population previously (Hammond et al., 2011; Supplemental Figure 3).

**Global Analyses of *B. rapa* Gene Expression Data Using Ca and Mg Supply as Cofactors**

eQTLs are genetic regions associated with variations in gene expression among individuals (Kliebenstein, 2009). This variation can arise due to sequence polymorphisms in target genes, their cis-regulatory (proximal) regions or trans-regulatory (distal) genes, leading to phenotypic differences. Identifying variation in gene expression within a segregating mapping population provides a powerful tool for resolving contributions and interactions within functional gene regulatory networks (Druka et al., 2010). We quantified gene expression levels for the 135k probe sets on the *Brassica* *Exon* 1.0 ST Affymetrix array in 85 lines of the BraIRRI mapping population under the LL, LH, and HL treatments. Since the array was designed to all *Brassica* sequence information available at the time, including *B. rapa, B. napus,* and *B. oleracea,* not all probe sequences correspond to unique *B. rapa* genomic locations (Love et al., 2010), and *B. rapa* gene models can be represented by more than one probe set. In total, 19,999 unique BRAD gene models were represented by probe sets on the array. Thus, of the 35,256 significant (log of the odds [LOD] score > 3.17) differential signals, 9910 had physical locations in the *B. rapa* genome (1895 cis and 8015 trans) and therefore were designated as eQTLs (Figure 3). The average LOD score for cis-eQTLs was 19.89 ± 0.97 (SE), and for trans-eQTLs the average LOD score was 5.85 ± 0.13 (SE), consistent with previous eQTL studies (Hammond et al., 2011). The heritability of gene expression and
The direction (positive or negative) of the additive effect (of the IMB211 allele) for an eQTL was the same under all treatment conditions for most of the significant eQTLs associated with B. rapa gene models (7419); therefore, these were designated as nonresponsive. The remaining eQTLs associated with B. rapa gene models were designated as Ca-responsive (high Ca/low Mg; 971), Mg-responsive (low Ca/high Mg; 972), or Ca- and Mg-responsive (548), based on whether the direction of the additive effect of an eQTL changed under high Ca supply only, under high Mg supply only, or when both Ca and Mg supply were altered, respectively. Among the 1895 cis-eQTLs, 23 were Ca-responsive, 6 were Mg-responsive, and 8 were both Ca- and Mg-responsive. Among the 8015 trans-eQTLs, 948 were Ca-responsive, 966 were Mg-responsive, and 540 were both Ca- and Mg-responsive. When the physical positions of eQTLs are plotted as a function of their genetic positions, a cis-diagonal can be seen clearly among the nonresponsive group. Among the 25,346 eQTLs lacking a physical position in the genome, 3615 were Ca-responsive, 3139 were Mg-responsive, and 2490 were both Ca- and Mg-responsive.

Hot spots of eQTLs can be seen when a mapping interval GEM has more eQTLs associated with it than would be expected by chance (Figure 4). Nonresponsive eQTL hot spots share a number of genomic locations with phosphorus-responsive hot spots identified.

**Figure 2.** Shoot Ca Concentrations of Three Genotypes of B. rapa (IMB211, R500, and R-o-18) as Functions of \([\text{Ca}]_{\text{ext}}\) and \([\text{Mg}]_{\text{ext}}\).

Plants were grown in compost for 18 d. Data are means of three replicates analyzed using a genotype/\([\text{Ca}]_{\text{ext}} \times [\text{Mg}]_{\text{ext}}\) treatment structure within an ANOVA. In (A) to (C), there are six \([\text{Mg}]_{\text{ext}}\) treatment levels per \([\text{Ca}]_{\text{ext}}\) treatment level; these are indicated as numerical symbols for each data point. Similarly, in (D) to (F), there are six \([\text{Ca}]_{\text{ext}}\) treatment levels per \([\text{Mg}]_{\text{ext}}\) treatment level. (G) to (I) are ternary plots showing interpolated data for all 36 Ca \(\times\) Mg treatment combinations. Error terms (LSD values; \(P = 0.05\)) are as follows: 689, 1689, and 4136 mg/kg for genotype, \([\text{Ca}]_{\text{ext}}\) or \([\text{Mg}]_{\text{ext}}\), and genotype/\([\text{Ca}]_{\text{ext}} \times [\text{Mg}]_{\text{ext}}\) terms, respectively.
In Silico Phenotyping of eQTL Orthologs Using a Large Arabidopsis Ionomics Database

In silico phenotyping using the Purdue Ionomics Information Management System (PiILMS; Baxter et al., 2007) database of Arabidopsis leaf ionomic profiles was performed to identify gene targets associated with eQTLs in B. rapa that also had shoot-Ca phenotypes associated with mutations in the orthologous gene in Arabidopsis. Identifying Arabidopsis genotypes with mutations in known genes that have a shoot-Ca phenotype and are associated with a significant eQTL in B. rapa provides an efficient pipeline to select eQTLs for further study. We identified 39 mutant genotypes in the PiILMS database with consistent and robust altered shoot-Ca phenotypes (i.e., which had increased [z score ≥ 3] or decreased [z score ≤ -3] leaf-Ca relative to a control plant in ≥50% of the samples for an individual mutant genotype). Of these, 24 genotypes had decreased leaf-Ca and 15 had increased leaf-Ca (Supplemental Data Set 4). These 39 genotypes have 31 unique Arabidopsis Genome Initiative (AGI) code tags. These AGI codes include four potential Ca2+ transporters: ACA8, an autoinhibited Ca2+-ATPase (Geisler et al., 2000); ECA4, a P2A-type Ca2+-ATPase located in the endoplasmic reticulum (Mills et al., 2008); and two Ca2+/-H+ antiporters, CAX1 and CAX3 (Shigaki and Hirschi, 2006; Conn et al., 2011). They also include a cation efflux protein of the cation diffusion facilitator family, MTP5 (Gustin et al., 2011).

The 31 AGI codes mapped to 62 probe sets on the Brassica Exon 1.0 ST Affymetrix array (Supplemental Data Set 5), of which 21 probe sets were associated with one or more significant eQTLs, for a total of 27 eQTLs. Of these, six eQTLs were designated as responsive: two Ca-responsive, three Mg-responsive, and one Ca- and Mg-responsive. Among the responsive eQTLs, two were associated with ACA8, two with CAX1, and two with an Arabidopsis FLOWERING LOCUS F MADS box protein (AT5G10140). Among the nonresponsive eQTLs, two were associated with CAX3 and one each with IAR1 (At1g68100), a member of the ZIP metal ion transporter family, and ECA4. CYTOKININ31, a blue light photoreceptor (AT1G04400), was associated with four nonresponsive eQTLs and has been implicated in the regulation of circadian oscillations in the concentration of cytosolic Ca2+ in response to blue light (Xu et al., 2007). The functional significance of this association is not known.

Ideally, data from a complete set of homozygous knockout mutants and overexpression lines of Arabidopsis are required to test the specificity of in silico phenotyping, but these data are not yet available. This study, which is thus biased toward transporter mutants, shows that 16 out of the 37 Brassica genes (43%) with a mapped physical location linked to an AGI code tag for shoot-Ca are linked to an eQTL. Similarly, 17 out of 32 Brassica genes (53%) linked to an AGI code tag for shoot-Mg are linked to an eQTL. A similar analysis for Fe (34%) and Zn (32%) shows these proportions to be lower, despite many mutants having multiple phenotypes. For example, 32 and 23% of Arabidopsis mutants with a phenotype of z > 3 or z < -3 for shoot-Ca concentration also have significantly altered leaf Fe and Zn concentration phenotypes, respectively.

Characterizing BraA.CAX1a Mutants

We had previously identified three missense mutations in the Ca2+ transporter BraA.CAX1a by TILLING (Ó Lochartaín et al., 2011).
Therefore, we conducted further selfing, genotyping, and characterization of these lines, based on the identification of a significant eQTL associated with BraA.CAX1a under high-Mg, low-Ca conditions and the subsequent in silico identification of a Ca phenotype associated with the Arabidopsis ortholog of this gene (Supplemental Data Sets 4 to 7). The characterization of this gene in B. rapa demonstrates the potential of the Brassica-Arabidopsis-Brassica workflow but was also guided by prior studies in Arabidopsis (Hirschi et al., 1996; Pittman and Hirschi, 2001; Shigaki et al., 2001, 2002; Català et al., 2003; Cheng et al., 2003, 2005).

A shoot-Ca phenotype in B. rapa was observed for BraA.CAX1a (Bra017134) mutants, which corresponded to an eQTL and an Arabidopsis phenotype in the ionomics database. To isolate lines with mutations in the BraA.CAX1a gene in B. rapa, a TILLING approach was adopted previously. A 1.5-kb fragment including the transcriptional start site and the first exon was used as the target. In total, 20 mutations were identified comprising 10 missense, 1 nonsense, 2 silent, and 7 noncoding changes (Ó Lochlainn et al., 2011; Supplemental Figure 4). Lines with missense and nonsense (stop) mutations were backcrossed to the R-o-18 parent line of the TILLING population, and homozygous and segregating wild-type lines for each mutation were isolated in the BC1S1 population. Of these lines, three missense alleles designated BraA.cax1a-4 (A-to-T change at amino acid 77), BraA.cax1a-7 (R-to-K change at amino acid 44), and BraA.cax1a-12 (P-to-S change at amino acid 56) were characterized in BC1S2.

Visibly different phenotypes were observed in the homozygous mutants BraA.cax1a-4 and BraA.cax1a-7 compared with their segregating wild types (Figure 5). Homozygous plants of the BraA.cax1a-4 and BraA.cax1a-7 lines had paler/yellow leaves at both 4 and 5 weeks old (Figure 5). This phenotype was seen in expanding and fully expanded leaves in BraA.cax1a-7 at both time points. However, in BraA.cax1a-4, expanding leaves were pale at weeks 4 and 5, but fully expanded leaves were only pale at week 5. The leaves of BraA.cax1a-12 appeared similar to those of the segregating wild-type and R-o-18 plants at both time points. To quantify this phenotype, leaf chlorophyll concentration was measured indirectly at two time points using a single-photon avalanche diode (SPAD) meter. Leaf chlorophyll concentration was lower in both the expanding and fully expanded leaves of BraA.cax1a-4 and its segregating wild type compared with the wild-type value (26,036 versus 34,356 mg/kg; P = 0.05, n = 3).
There was no significant difference in shoot-Ca in BraA.cax1a-4 and its segregating wild type. The only significant shoot-Mg phenotype was seen in BraA.cax1a-12, which had a higher shoot-Mg in the mutant compared with the wild type (18,733 versus 12,544 mg/kg; \( P = 0.04, n = 3 \)). There were no mutant phenotypes observed for shoot Zn, Fe, or K concentration (Table 1).

Further studies are needed to resolve the variation in shoot-Ca phenotypes observed among *B. rapa* missense mutants of *BraA.CAX1a*, especially given the relative complexity of the *B. rapa* genome compared with *Arabidopsis*. However, even in the simpler genome of *Arabidopsis*, shoot-Ca phenotypes are not always consistent in lines containing loss-of-function mutations in *CAX1* (White, 2014). For example, Català et al. (2003) reported *Arabidopsis cax1* mutants having lower shoot-Ca than wild-type plants, and Conn et al. (2011) reported a positive correlation between *CAX1* expression and shoot-Ca in 15 ecotypes of *Arabidopsis*. By contrast, other groups have not reported reduced shoot-Ca in *At-cax1* mutants (Cheng et al., 2003, 2005; Conn et al., 2011), possibly due to functional compensation by *CAX3* (Cheng et al., 2005; Conn et al., 2011), although *CAX3* expression does not correlate with shoot-Ca in the same way as that of *CAX1* (Conn et al., 2011). Consistent with this interpretation, the shoot-Ca of *cax1 cax3* double mutants is often significantly lower than that of wild-type plants, and impaired gas exchange has been observed (Cheng et al., 2005; Conn et al., 2011, 2012). No previous reports show an *Arabidopsis cax1* loss-of-function mutant having increased shoot-Ca. However, there are numerous reports of altered tissue

### Table 1. Mean Shoot Mineral Compositions of *BraA.cax1a* Mutants and Their Segregating Wild Types

<table>
<thead>
<tr>
<th>Line</th>
<th>Shoot-Ca (mg/kg)</th>
<th>SE</th>
<th>Shoot-Mg (mg/kg)</th>
<th>SE</th>
<th>Shoot-Zn (mg/kg)</th>
<th>SE</th>
<th>Shoot-Fe (mg/kg)</th>
<th>SE</th>
<th>Shoot-K (mg/kg)</th>
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<td>44,534</td>
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<td>55.2</td>
<td>8.8</td>
<td>54,875</td>
<td>12,435</td>
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<td>1,392</td>
<td>12,544</td>
<td>633</td>
<td>62.3</td>
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<td>75.2</td>
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<td>8,801</td>
</tr>
<tr>
<td><em>BraA.CAX1a-4</em> WT</td>
<td>44,379</td>
<td>9,704</td>
<td>21,908</td>
<td>6,247</td>
<td>68.8</td>
<td>10.8</td>
<td>63.5</td>
<td>13.4</td>
<td>54,225</td>
<td>14,634</td>
</tr>
<tr>
<td>P</td>
<td>0.327</td>
<td>0.279</td>
<td>0.680</td>
<td></td>
<td>0.752</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>0.447</td>
</tr>
</tbody>
</table>

Shoot samples (\( n = 3 \)) were analyzed by inductively coupled plasma mass spectrometry. \( P \) values were calculated using a two-sided Student’s t test.
Ca concentrations arising from the overexpression of genes encoding truncated versions of At-CAX proteins lacking autoinhibitory domains (White, 2014). For example, unregulated expression of sAt-CAX1, a modified At-CAX2 gene (sAt-CAX2), or At-CAX4 has been shown to increase Ca concentrations in edible portions of several crops, including carrot, lettuce (Lactuca sativa), tomato, and potato (Park et al., 2004, 2005a, 2005b, 2009; Morris et al., 2008).

Comparative genomic studies are now feasible between plant species, including those with complex genomes such as B. rapa (King, 2013). Next-generation sequencing has facilitated the identification of paralogous and orthologous genes, resulting from genome segmentation and polyploidy events in plant genomes, even within large gene families (Cheng et al., 2012). In B. rapa, there are two paralogous copies of BraA.CAX1 but only a single version of BraA.CAX3. It will be necessary to isolate mutants of these individually and in combination to test for redundancy between gene family members and between paralogous copies to establish definitive roles for these proteins in Ca transport and storage. The three independent missense allelic variants in BraA.CAX1 characterized here target different amino acids upstream of the N-terminal autoinhibitory domain (Supplemental Figure 4; Pittman and Hirsch, 2001), offering insights into other residues critical for the regulation and function of this protein.

TILLING is a powerful approach for generating an allelic series of mutations within genes to study more subtle alterations in protein expression and structure. The development of these resources in crop species will assist in the functional characterization of genes and their products beyond model systems and shed new light on gene function. Given that Brassica species are consumed widely as vegetable crops and are possible target crops for increasing intake of important nutrients such as Ca (Broadley and White, 2010), the knowledge and resources generated here will be of use in future strategies for the biofortification of edible crops. It is important to stress that the Brassica-Arabidopsis-Brassica workflow is currently dependent on (1) extensive prior annotation and characterization in Arabidopsis and (2) the availability of leaf Ca concentration data from the Arabidopsis ionomics database, which is currently weighted toward well-characterized transporter mutants. A full set of phenotypic data from homozygous knockout mutants and overexpression lines of Arabidopsis would be the preferred in silico resource in the coming years to improve the workflow and identify novel genes where a priori information is not available.

METHODS

Responses of Brassica rapa to Altered [Ca]ext and [Mg]ext

Two parental lines of the B. rapa (2n = 2x = 20; A genome) BraIRRI population, IMB211 and R500 (Iniguez-Luy et al., 2009; Hammond et al., 2011), and the R-o-18 reference parent of the B. rapa TILLING population (Stephenson et al., 2010) were grown in a compost substrate in 9 × 9 × 8-cm pots (Desch Plantpak) in a controlled environment. The controlled environment was a walk-in high-specification growth room (Weiss-Gallenkamp) at Wellesbourne, UK, set to a 16-h photoperiod using metal halide lamps (Philips Sun-T Pia Greenpower; Philips Electricals) for 18 h/d. A combination of Ca and Mg applications was made to give four treatments (LL, LH, HL, and HH): the high concentrations were 3.5 g/L compost (24 mM) CaCl2 and 3.04 g/L (15 mM) MgCl2, and the low concentrations were 0.44 g/L (3 mM) CaCl2 and 0.2 g/L (1 mM) MgCl2. Twelve pots of each RIL and 36 pots of each parent line were sown per treatment using a randomized block design. Fully expanded leaves from six plants per RIL and 18 plants per parent were sampled 18 d after sowing, bulked (in pools of six for parent lines), and snap-frozen in liquid N2. In total, 268 biological samples were obtained, comprising three replicates for each parent line at each treatment (3 replicates × 4 treatments × 2 parents = 24) and samples from 85, 81, 65, and 13 unique samples of RIL from the LL, LH, HL, and HH treatments, respectively.

RNA Extraction and Array Analysis

For each sample, total RNA was extracted using a modified TRIzol extraction protocol (Hammond et al., 2006) and purified using RNA Cleanup for RNeasy columns with on-column DNase digestion (Qiagen). Total RNA was labeled using the Affymetrix WT Labeling Kit and hybridized to the Affymetrix Brassica Exon 1.0 ST array (Love et al., 2010) according to the
manufacturer’s instructions at two service providers: Nottingham Arabidopsis Stock Centre (University of Nottingham) and Source Bioscience. The total number of RNA samples processed on the Affymetrix Brassica Exon 1.0 ST array (and therefore *.cel output files) was 292. These samples were made up of the 268 biological samples described above plus 24 technical replicates, comprising 12 samples from parent lines grown under LL and LH treatments, 8 RNA samples from selected BraIRRI lines from the LL treatment, and 4 RNA samples from selected BraIRRI lines from the LH treatment. For the technical replicates, the RNA sample was split and hybridized to arrays independently by the two service providers. All data were imported into GeneSpring GX (version 11.0.2; Agilent Technologies) and normalized using the robust multiarray average (RMA) algorithm (Bolstad et al., 2003; Irizarry et al., 2003a, 2003b). All data have been deposited in the National Center for Biotechnology Information’s Gene Expression Omnibus (Edgar et al., 2002) and are accessible through GEO series accession number GSE44185. The normalized signal value for an individual probe set from an array was standardized to the median signal value for that probe set across all arrays. Comparisons between technical replicates from different service providers identified 229 differentially (2-fold; P < 0.01, Benjamini-Hochberg false discovery-corrected P value based on paired f tests between technical replicates) expressed probe sets across all technical replicates, equivalent to 0.17% of probe sets. The average correlation coefficient of raw expression values between technical pairs was 0.958 ± 0.002 (tสถ; n = 24).

Development of a Linkage Map Using GEMs

A set of 134 GEMs was selected for the BraIRRI_05 subpopulation using an approach modified from a previous study (Hammond et al., 2011). The approach of Hammond et al. (2011) was based on the Agilent Brassica 95k 60-mer arrays with a single probe per gene model. By contrast, probe sets on the Affymetrix Exon 1.0 ST array are derived from multiple 25-mer probes (Love et al., 2010). Therefore, because signal values are averaged across multiple probes, polymorphisms affecting the signal values of individual probes are likely to be masked. To overcome this constraint, exon probe sets, consisting of fewer probes, were used to identify polymorphic GEMs between the two parent lines. First, exon-level *.cel files for the two parents under all four treatments (i.e., IMB211 [n = 18 (3 replicates × 4 treatments) + 6 technical replicates] and R500 [n = 18]) were normalized using the RMA normalization algorithm (Bolstad et al., 2003; Irizarry et al., 2003a, 2003b), and the signal value for each exon probe set on an array was standardized to the median signal value for that exon probe set across all arrays. Second, exon probe set signal values were ranked according to segregation differences between parents based on P values from a two-tailed Student’s t test. A subset of 3272 exon probe sets, representing 2358 unique unigenes, were selected for exploration as suitable GEMs based on a segregation threshold of P < 2.1 × 10−8 between parents (Supplemental Data Set 3).

For mapping using GEMs, a reiterative procedure was followed using JoinMap4.1 (Kyazma; van Ooijen, 2006) with a combination of maximum likelihood mapping (MLM) and regression mapping (RM). The midpoint between mean parental exon probe set signal values was used to provisionally classify offspring as IMB211 (a allele) or R500 (b allele). A distinctness score was calculated for each exon probe set on the basis of the SD of the a-class offspring and the SD of the b-class offspring. High values of distinctness should give more accurate allele calling and might reflect sequence-based polymorphisms. The 350 GEMs with the highest distinctness scores were used for initial grouping using the “independence LOD” option, ranging from LOD 2 to 15 in 2-LOD steps. Generally, the final LOD stringency selected was between LOD 3 and LOD 5 based on visual inspection of linkage groups. For initial linkage group maps, default settings were used in MLM and RM (Haldane mapping function; “jump” value of 5). The robustness of GEMs was tested by comparing results from MLM and RM. Thus, an initial set of GEMs was generated through iterative removal of GEMs by examining “fit and stress” parameters for MLM and by visual inspection of graphical genotypes for MLM and RM. Miscalled GEMs were removed (e.g., double recombination events in individual lines within small genetic distances [1 to 5 cM]). Further GEM removal was conducted to give a spaced genetic map of 134 GEMs for 85 RILs with intervals of between 5 and 10 cM for the subsequent eQTL analysis.

Global Analyses of B. rapa Expression Data Using Ca and Mg Supply Treatments as Cofactors

Gene expression data from three treatments (n = 273 *.cel files) were again normalized using RMA. These data represented (1) 105 LL *.cel files from 85 RILs plus 8 technical replicates and 12 parent files comprising biological (n = 3) and technical (n = 3) replicates from two parents; (2) 97 LH *.cel files from 81 RILs plus 4 technical replicates and 12 parent files as above; and (3) 71 HL *.cel files from 65 RILs plus biological (n = 3) replicates from two parents. The HH treatment was excluded due to low sample numbers. Log2-normalized signal values were exported as a data matrix with 273 columns (RNA samples) and 135,264 (135k) rows (array probe sets). Custom GenStat scripts (available on request from the authors) were used to batch-process the log2-normalized signal values in groups of up to 1500 genes. Genotypic and environmental variance components were estimated for each probe set using a residual maximum likelihood procedure (Patterson and Thompson, 1971; Robinson, 1987; Hammond et al., 2011), where the genotypic variance component is the effect of line within the population and the environment variance component is the treatment effect of altered Ca and Mg supply. Thus, the variance components model included gene expression as the variate, treatment Vtreat) as a fixed model term, and line (line × treatment) [Vline + (Vline × Vtreat)] and a residual (Vres) variance random model term. For each probe set, the treatment effect was defined as follows:

\[
\text{Treatment effect } = \left( \frac{V\text{line}}{V\text{treat}} + V\text{line} + (V\text{line} + V\text{line}) \right) - \left( \frac{V\text{line}}{V\text{line} + V\text{line}} \right)
\]

The heritability of the expression of each probe set was interpreted as a broad-sense mean line heritability based on Cullis et al. (2006). A variance–covariance matrix (\(\Sigma\)) was derived from the residual maximum likelihood for each probe set in turn, which is the line × line matrix yielding an expression variance term for each line (\(V\text{line}\)). This matrix thereby yields a diagonal (\(\text{dpev}\)) of variance terms across all lines. Subsequently, the mean of this diagonal is used [mean (\(\text{dpev}\))], for each gene in turn, to estimate the heritability of gene expression across the population (\(H^2\)) as follows:

\[
H^2 = 1 - \left( \frac{\text{mean}(\text{dpev})}{V\text{line}} \right)
\]

Identifying and Positioning Significant eQTLs

Data for parental lines were removed from the above (273 *.cel files) data set to generate a data matrix of 243 columns and 135k rows (probe sets). These data were used to identify eQTLs using a two-stage process. First, the GenStat procedure QIBDPROBABILITIES (Boer and Thissen, 2009) was used to calculate a set of genetic predictors at the marker positions on the GEMs genetic map (BraIRRI_05_2012a) for 85 RILs. A parallel regression model was used to identify possible QTL effects across 1500 probe sets simultaneously, by testing for a combined QTL × environment effect, where QTL indicates a genetic predictor used as a covariate. This model was fitted for each genetic predictor in turn. Second, for probe sets with a significant (LOD score > 3.3, P < 0.0005) result in this first step, simple interval mapping was performed using QTL procedures in
GenStat using genetic predictors within the mixed model framework and a LOD threshold > 3.17 (Boer et al., 2007). Briefly, a mixed model was fitted with fixed environmental effects and QTL × environment interaction for each genetic predictor in turn and random terms comprising $V_{\text{gen}}$, $V_{\text{pve}}$, $V_{\text{vpe}}$, and $V_{\text{ve}}$. A compound symmetry structure was used to account for genetic covariance across environments. The additive allelic effect was defined as half of the difference in probe set expression between parental alleles at each marker and reported with respect to the IMB211 (a allele).

Individual probes sets on the Affymetrix Brassica Exon 1.0 ST array were aligned to the B. rapa Chifu-401 genome sequence (version 1.0; 255.9 Mb, representing 90% of the assembled sequences) (Wang et al., 2011). Probe sets were assigned to specific chromosome sequence coordinates when >70% of probes within the probe set aligned with a match of ≥98% within the genome sequence. A cis-eQTL was defined when the physical position of the probe set and its associated GEM were in the same region (i.e., at the midpoint between two GEMS or between a GEM and the end of the chromosome), and the QTL hot spots were defined when the number of eQTLs associated with a GEM exceeded an empirical threshold (the upper 99% confidence interval for the Poisson distribution, assuming an equal distribution of GEMS and eQTLs). Genesect (Virtual Plant 1.3; Katari et al., 2010) was used to test for overlaps in the genes associated with eQTL hot spots identified both here and by Hammond et al. (2011), based on z scores.

Four nonredundant gene lists were defined according to whether the direction (positive, the IMB211 allele resulted in a greater expression value of the gene; or negative, the IMB211 allele resulted in a lower expression value of the gene) of the additive effect of the eQTL was the same under all treatment conditions (nonresponsive) or whether it changed under altered Ca supply only (Ca-responsive), under altered Mg supply only (Mg-responsive), or when both Ca and Mg supply were altered (Ca- and Mg-responsive).

In Silico Phenotyping of eQTLs Using Arabidopsis thaliana Databases

Probe Set Annotation and Comparative Genomics

Each probe set on the Affymetrix Brassica Exon 1.0 ST array was mapped to the B. rapa gene lists from the BRAD database (version 1.2; plantgdb.org/BrGDB/) (Cheng et al., 2011). A match between a probe set and a gene was defined if at least one probe in that probe set had an exact, full-length, ungapped alignment to that gene. The probes were aligned using TimeLogic DeCypher Tera-BLASTN. Brassica genes most similar in sequence to a particular Arabidopsis gene were found by aligning Brassica protein sequences (BRAD version 1.2) to Arabidopsis protein sequences from TAIR (release 10; http://www.arabidopsis.org) using TimeLogic DeCypher Tera-BLASTN (e value cutoff = 10e-4). For each Brassica protein sequence, a single match was defined as the Arabidopsis sequence aligning with the highest bit score. Gene Ontology enrichment was performed using the Gene Ontology category analysis tool in Gene Spring GX, using a hypergeometric test, $P < 0.05$ with a Benjamin-Yekutieli correction for multiple testing (Benjamini and Yekutieli, 2001).

Comparative Analysis of an Arabidopsis Ionomics Data Set

An ionomics database of 152,601 unique samples (“tubes”) representing shoot tissue concentrations of Ca, Mg, and 21 other elements in Arabidopsis (hosted by PiMiS [Baxter et al., 2007]; downloaded October 26, 2011) was used for in silico phenotyping. Of these samples, 7992 had an altered shoot-Ca phenotype as defined by a z score of less than −3.0 (decreased shoot-Ca) or greater than +3.0 (increased shoot-Ca) compared with a control: 3048 and 4944 samples had increased and decreased shoot-Ca, respectively. Among these 7992 samples, 3272 were of mutant genotypes/lines tagged with one or more Arabidopsis gene identifiers; 1154 and 2118 had increased and decreased shoot-Ca, respectively (Supplemental Data Set 6). These 3272 samples represented 682 Arabidopsis genotypes/lines where one or more samples had a leaf-Ca phenotype, of which 562 were single gene mutants and the remainder were multiple mutants (Supplemental Data Set 7). Among these 682 Arabidopsis genotypes/lines, 39 had at least 50% of their samples in the PiMiS database showing altered shoot-Ca, demonstrating a consistent shoot-Ca phenotype; 24 and 15 genotypes/lines had increased and decreased shoot-Ca, respectively (Supplemental Data Set 4). There were 31 unique Arabidopsis gene identifiers associated with these 39 genotypes/lines (Supplemental Data Set 5).

Characterizing Bra.A.CAX1a Mutants

Mutations in the Bra.A.CAX1a (Bra017134) gene were identified in a B. rapa TILLING population as described by ŌLochlainn et al. (2011) using the RevGenUK service (http://revgenuk.jic.ac.uk). A 1.5-kb fragment, including the transcriptional start site and the first exon, was used as the target for TILLING. In total, 20 mutations were identified, including 10 missense, 1 nonsense, 2 silent, and 7 noncoding changes. M3 lines with missense and nonsense mutations were back-crossed to the R-o-18 parent line of the TILLING population. The BC1 plants were grown and individual plants genotyped using high-resolution melt analysis (ŌLochlainn et al., 2011), modified to include Meltdoctor HRM Master Mix (Applied Biosystems) according to the manufacturer’s instructions. Plants heterozygous for their mutation were selfed, and 20 individual BC1S1, BC1S2, and BC1S3 plants were grown and genotyped. Homozygous mutants of three missense alleles designated Bra.A.cax1a-4 (A-to-T change at amino acid 77), Bra.A.cax1a-7 (R-to-K change at amino acid 44), and Bra.A.cax1a-12 (P-to-S change at amino acid 58) and their segregating wild-type lines were identified and selfed to BC1S2 for phenotyping. Homozygous mutants and the corresponding wild type from segregation for each of Bra.A.cax1a-4, Bra.A.cax1a-7, and Bra.A.cax1a-12, and the original R-o-18 parent line, were phenotyped (i.e., seven lines in total). Plants were grown in a greenhouse at Sutton Bonington, UK, in July and August 2012. Plants were grown in 13-cm-diameter pots containing 1 liter of Levinton M2 compost (J. Arthur Bowers). Supplementary lighting was provided as described above, and plants were watered with tap water as required. Total leaf Ca, Mg, and chlorophyll concentrations were determined on a single fully expanded leaf (true leaf number 4 or 5) after 5 weeks from triplicate plants of each line. Total leaf Ca and Mg concentrations were determined by a commercial foliar analysis laboratory (Yara Analytical Services, Lançrop Laboratories, Yara UK) using inductively coupled plasma atomic emission spectrometry. Leaf chlorophyll concentration was measured indirectly on an expanding leaf and a fully expanded leaf using a SPAD meter (SPAD-502; Konica Minolta); three measurements were taken per leaf at 4 and 5 weeks. All data were analyzed using GenStat.

Accession Numbers

Sequence data from this article can be found in the Arabidopsis Genome Initiative or GenBank/EMBL databases under the following accession numbers: Bra.A.CAX1a, Bra017134; and At-CAX1, At2g38170.

Supplemental Data

The following materials are available in the online version of this article.

Supplemental Figure 1. Target and Measured [Ca]ext and [Mg]ext in Compost.

Supplemental Figure 2. Leaf Mg Concentrations of Three Genotypes of B. rapa (IMB211, R500, and R-o-18) as Functions of External CaCl2 and MgCl2 Supply.
Supplemental Figure 3. Summary of Heritability (H) and Treatment Effect Variance Components for 135k Brassica Array Probe Sets.

Supplemental Figure 4. Putative Structure of BraA.CAX1a.

Supplemental Figure 5. Single-Photon Avalanche Diode (SPAD) Measurements of Expanding and Fully Expanded Leaves of Three Independent Homozygous BraA.cax1a Lines and Their Segregating Wild Types.

Supplemental Table 1. Analysis of Nonresponsive eQTL Hot Spots.

Supplemental Data Set 1. All Raw Primary Biomass and Tissue Concentration Data for Plants Grown under 36 [Ca]ext × [Mg]ext Conditions.

Supplemental Data Set 2. Selection of 3272 GEMs Based on Exon-Level Probe Sets.

Supplemental Data Set 3. BraIRRI_05_2012a Map from 134 GEMs.

Supplemental Data Set 4. Arabidopsis Genotypes with Altered Leaf-Ca in ≥50% of Their Ionomics Samples.

Supplemental Data Set 5. B. rapa Probe Sets and eQTLs Linked to 31 Arabidopsis Genes Whose Mutants Have Altered Shoot-Ca Phenotypes.

Supplemental Data Set 6. Ionomics Data for 3272 Samples with Altered Shoot-Ca.

Supplemental Data Set 7. Arabidopsis Genotypes/Lines with Altered Shoot-Ca in the Ionomics Database (n = 682).

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AUTHOR CONTRIBUTIONS

N.S.G., J.P.H., G.J.K., P.J.W., and M.R.B. conceived and designed the project. J.P.H. and H.C.B. performed growth-response experiments. N.S.G., J.P.H., S.Ó.L., and M.R.B. performed eQTL experiments. S.M. conducted GEM and mapping analyses. N.S.G., S.Ó.L., B.B., and J.J.R. performed TILLING experiments. N.S.G., J.P.H., A.L., C.J.R., S.W., P.W.C.C., L.X.D., and M.R.B. conducted data analyses. All authors contributed to the writing and editing of the article. All authors have read and approved the final article.

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Genetical and Comparative Genomics of *Brassica* under Altered Ca Supply Identifies *Arabidopsis* Ca-Transporter Orthologs

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