

RESEARCH PAPER

Prioritizing quantitative trait loci for root system architecture in tetraploid wheat

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Received 29 September 2015; Accepted 18 January 2016

Editor: Adam Price, University of Aberdeen

Abstract

Optimization of root system architecture (RSA) traits is an important objective for modern wheat breeding. Linkage and association mapping for RSA in two recombinant inbred line populations and one association mapping panel of 183 elite durum wheat (Triticum turgidum L. var. durum Desf.) accessions evaluated as seedlings grown on filter paper/polycarbonate screening plates revealed 20 clusters of quantitative trait loci (QTLs) for root length and number, as well as 30 QTLs for root growth angle (RGA). Divergent RGA phenotypes observed by seminal root screening were validated by root phenotyping of field-grown adult plants. QTLs were mapped on a high-density tetraploid consensus map based on transcriptassociated Illumina 90K single nucleotide polymorphisms (SNPs) developed for bread and durum wheat, thus allowing for an accurate cross-referencing of RSA QTLs between durum and bread wheat. Among the main QTL clusters for root length and number highlighted in this study, 15 overlapped with QTLs for multiple RSA traits reported in bread wheat, while out of 30 QTLs for RGA, only six showed co-location with previously reported QTLs in wheat. Based on their relative additive effects/significance, allelic distribution in the association mapping panel, and co-location with QTLs for grain weight and grain yield, the RSA QTLs have been prioritized in terms of breeding value. Three major QTL clusters for root length and number (RSA QTL cluster 5#, RSA QTL cluster 6#, and RSA QTL cluster 12#) and nine RGA QTL clusters (QRGA.ubo-2A.1, QRGA.ubo-2A.3, QRGA.ubo-2B.2/2B.3, QRGA.ubo-4B.4, QRGA.ubo-6A.1, QRGA.ubo-6A.2, QRGA.ubo-7A.1, QRGA. ubo-7A.2, and QRGA.ubo-7B) appear particularly valuable for further characterization towards a possible implementation of breeding applications in marker-assisted selection and/or cloning of the causal genes underlying the QTLs.

Key words: Association mapping, drought stress, germplasm collection, GWAS, grain yield, meta-QTLs, root growth angle, rooting depth, root system architecture, seminal root.

Introduction

Root system architecture (RSA) plays a pivotal role in crop performance, particularly for cultivation under nonoptimal water and nutritional supply conditions (Ludlow and Muchow, 1990; de Dorlodot *et al.*, 2007; Paez-Garcia *et al.*, 2015). Based on the expected climate changes and the declining availability of water and fertilizers, enhancing the

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genetic capacity of the plant to acquire soil resources is a primary target to increase crop productivity and yield stability (Hawkesford et al., 2014; Mickelbart et al., 2015). In the past decade, RSA has received increasing attention in cereals (Hochholdinger and Tuberosa, 2009; Wasson et al. 2012, 2014; Bishopp and Lynch, 2015), leading to the development of detailed RSA ideotypes (King et al., 2003; Lynch, 2013; Meister et al., 2014). In rice, a narrow and deep root ideotype for enhancing drought resistance has been successfully pursued based on direct field observation of root distribution (Steele et al., 2013; Uga et al., 2013) or root growth angle (RGA) measurements in rhizotrons (Kitomi et al., 2015). In sorghum, stay-green genotypes have contributed additional evidence for the positive role on yield of narrow RGA quantitative trait loci (QTLs) under drought conditions (Borrell et al., 2014). RGA is also of paramount importance for the acquisition of phosphorus, a low-mobility nutrient usually more abundant in the upper soil layer (Miguel et al., 2015).

Among cereals, wheat is prevalently grown under rainfed conditions in regions where drought stress is the major environmental factor limiting productivity. Accordingly, in at least 60 million rainfed hectares, grain yield of wheat was only 10–50% of that reached under irrigation (Fleury et al., 2010; Langridge and Reynolds, 2015). Drought can affect wheat at all vegetative stages, mainly from flowering to grain filling in Mediterranean environments. Breeding for enhanced water and nutrient uptake would therefore result in increased yield and yield stability, particularly under water-limited environments (Manschadi et al., 2010; Wasson et al., 2012; Christopher et al., 2013; Lopes et al., 2014). Optimizing the anatomy and growth features of roots can significantly increase water-use efficiency (WUE; Richards and Passioura, 1989; Wasson et al., 2012) and/or moisture extraction from deep soil layers (Blum, 2009; Uga et al., 2013; Pinto and Reynolds, 2015). In wheat, lack of information on the effects of RSA QTLs on yield across water regimes has so far hindered the adoption of markerassisted selection for tailoring RSA, unlike in rice (Steele et al., 2013; Arai-Sanoh et al., 2014) and sorghum (Borrell et al., 2014).

Wheat shows two main root systems, namely the seminal (embryonal) roots and the nodal (crown or adventitious) roots (Chochois *et al.*, 2015). Seminal roots in cultivated wheat include one primary root, two pairs of symmetric roots, and, at times, a sixth central root. Nodal roots usually become visible when the fourth leaf emerges at the tillering stage (Esau, 1965). Seminal roots penetrate the soil earlier and more deeply than nodal roots and remain functional for the entire plant cycle, hence contributing to moisture extraction from deeper soil layers (Manschadi *et al.*, 2013).

Direct measurements of root length density across soil profiles showed that wheat growth on residual moisture greatly depends on roots that reach deep soil layers (Reynolds *et al.*, 2007; Acuna *et al.*, 2012; Hamada *et al.*, 2012; Manschadi *et al.*, 2013). Importantly, rooting depth has been related to the RGA of seminal roots as first reported by Oyanagi *et al.* (1993). More recently, this concept has been adopted for breeding purposes in rice (Uga *et al.*, 2015).

RGA is easily measured on seminal roots of seedlings (Sanguineti *et al.*, 2007; Richard *et al.*, 2015) or on the adventitious roots at the adult stage (Oyanagi *et al.*, 1993; Kato *et al.*, 2006). While the latter appears more relevant for crop performance, direct field-based RSA phenotyping of adult plants remains a labour-demanding undertaking, especially when experimental uniformity is required and experiments involve a large number of accessions (Mace *et al.*, 2012; Tuberosa, 2012). Conversely, RSA characterization at the seminal stage allows for an accurate, fast, and cheap evaluation of hundreds of accessions (Mace *et al.*, 2012). This notwithstanding, limited information is available on QTLs for seminal RSA in wheat (Canè *et al.*, 2014).

QTLs for RSA traits have been reported for diverse cereals (Tuberosa *et al.*, 2003; Hund *et al.*, 2011; Mace *et al.*, 2012; Christopher *et al.*, 2013; Courtois *et al.*, 2013; Acuna *et al.*, 2014; Borrell *et al.*, 2014). In rice, the identification of major QTLs for root depth has been instrumental in increasing adaptation to low water availability (Steele *et al.*, 2013) as well as for the positional cloning of *deeper rooting 1* (*DRO1*; Uga *et al.*, 2013).

Dissecting the genetic control of RSA traits is particularly important in tetraploid durum wheat (Triticum turgidum L. var. durum Desf.), a crop mainly grown under rainfed conditions and low water availability. In this study, linkage and association mapping were both used to (i) generate a comprehensive view of the QTLome (as defined in Salvi and Tuberosa, 2015) governing RSA traits in elite durum wheat at the seedling stage and (ii) investigate how these QTLs influence yield and overlap with similar QTLs in hexaploid wheat. A high-density tetraploid consensus map facilitated cross-referencing of QTLs from diverse materials and studies, hence allowing us to prioritize RSA QTLs for further studies. The genetic variation for RSA in the A and B wheat genomes is described, providing the necessary knowledge ultimately to fine-tune the expression of the RSA and model its expression based on genetic information.

Materials and methods

Plant material

Two recombinant inbred line (RIL) $F_{6:7}$ populations were developed by Produttori Sementi Bologna (Bologna, Italy) from Colosseo×Lloyd (Co×Ld; 176 RILs) and Meridiano×Claudio (Mr×Cl; 181 RILs).

Colosseo is an Italian cultivar released in 1990 with a high yield potential but poorly adaptated to Southern Mediterranean environments under high terminal drought and heat (as from the Italian National Network Trial annual reports, 1995–2005). Pedigree information indicates a direct origin from Creso (one of the founders of the modern germplasm, obtained from Italian and CIMMYT early Green Revolution materials). However, microsatellite data showed that Colosseo has chromosome segments directly introgressed from Mediterranean landraces (Maccaferri *et al.*, 2007). Lloyd (Cando/ Edmore) is a US cultivar well adapted to the relatively low-input conditions typical of the USA Northern Plains.

Meridiano is a medium to early maturing, widely adapted Italian cultivar derived from a complex cross between Italian, CIMMYT,

and Southwestern US materials (Simeto/WB881//Duilio/F21). Claudio, another Italian cultivar (CIMMYT selection/Durango// ISI938/Grazia), is renowned for its high yield stability across drought- and heat-stressed environments of Southern Europe (Spain, Italy, and Greece). More details are reported in Maccaferri *et al.* (2005, 2011) and De Vita *et al.* (2010).

The association panel, hereafter referred to as Unibo-DP (standing for 'UniBO Durum Panel'), includes 183 elite cultivars and lines from Mediterranean countries, the Southwestern USA, and Mexico (Maccaferri et al., 2011). Based on the characterization of simple sequence repeat (SSR) markers, the population structure of the Unibo-DP accessions herein considered appeared to be structured into five main subgroups representing the main breeding lineages present in the germplasm, identified by well-defined breeding ideotypes (and corresponding hallmark founders developed in different decades of breeding). Briefly, these subgroups correspond to: S1, ICARDA and Italian accessions for dryland areas from the native Syrian and North African germplasm (from Haurani and related landraces); S2, ICARDA accessions bred for temperate areas (from Cham 1); S3, Italian cultivars related to Valnova and Creso founders and subsequently bred with CIMMYT and Southwestern US accessions (Desert Durum[®]); S4, widely adapted early CIMMYT germplasm introduced to several Mediterranean countries (from Yavaros 79, Karim, Duilio); S5, more recent high yield potential CIMMYT germplasm (from Altar84). Details are reported in Maccaferri et al. (2011) and in Letta et al. (2013). Parental line Colosseo is grouped in S3 cluster, which harbours 38 accessions, 10 of which are derivatives of Creso, including Colosseo. Parental line Lloyd genome, belonging to the Northern Plains US germplasm, is not well represented in the Unibo-DP, except for the Southwestern US germplasm and recent Italian and French cultivars which inherited several portions of the genome from the US germplasm. Parental line Meridiano appeared mostly related to the S4 group, together with 60 other accessions from ICARDA, CIMMYT, and Mediterranean countries all related to the founder cultivars Yavaros C79, Karim, and Duilio (the same germplasm released at the CIMMYT and introduced in Tunisia and Italy). Parental line Claudio is a more diversified accession which loosely clusters with the CIMMYT-derived germplasm S4 and S5. More details are given in Maccaferri et al. (2011).

Phenotypic analysis

Evaluation at the seedling stage The mapping populations were characterized for RSA traits at the seedling stage using the protocol reported in Canè *et al.* (2014). Briefly, seeds were manually selected for uniformity, weighed, sterilized in a 1% sodium hypochlorite solution for 10′, rinsed in distilled water, and germinated in Petri dishes at 28 °C for 24 h.

Homogeneously sprouting seedlings were then grown in moist filter paper sheets in vertical black polycarbonate screening plates (42.5 × 38.5 cm). Seedlings were grown for 9 d at 22 °C (day)/18 °C (night) under a 16h light photoperiod and light intensity of 400 μ mol m⁻² s⁻¹ photosynthetically active radiation (PAR). The weight of the seeds used in the experiment, hereafter reported as 'initial thousand grain weight' (iTGW), was subjected to QTL analysis together with the RSA traits.

The experiments were conducted according to a randomized complete block design for each of the two populations and the Unibo-DP, with three independent replications (=experiments) grown consecutively in the same growth chamber. The experimental unit was equal to a 60 cm wide screening plate allotting eight seedlings of the same genotype, so that one screening plate corresponded to one genotype. To avoid border effects, root seminal architecture traits were measured from the six central seedlings only. Due to the high number of genotypes to be evaluated and considering the time needed for root preparation and acquisition of photographs of the roots, blocking was introduced to control for possible differences in growth rate. Twenty-five to 30 accessions were included in one block, corresponding to six blocks for each of the two populations and the Unibo-DP. Blocks corresponded to shelves in the growth chamber, positioned at the same distance from the floor under uniform light conditions.

Blocking was taken into account in the ANOVA, and linear adjustment for block effect (normalization) was carried out as necessary.

Root seminal architecture traits were measured on a single-plant basis: RGA (see Fig. 1A for an example) was obtained as the linear distance between the two most external roots of each plantlet at 3.5 cm from the seed tip and then converted to degrees; total number of roots (TRN); presence of the sixth seminal root (Rt6); and shoot length (SL). Additional RSA traits such as single root length, surface, diameter, and volume (PRL, PRS, PRD, and PRV, respectively) were measured on images using a digital camera with a shutter time set at 100.

Images were processed using SmartRoot[®] (Lobet *et al.*, 2011). Data from the primary roots were considered separately from those of the other roots. Lateral roots were counted on the primary root using a magnifying lens. After measuring SL, shoots and roots were cut and dried to determine the total dry weight for each replicate (SDW and RDW, respectively).

The root seminal architecture data for the DurumPanel accessions are reported in the DArT- and SSR-based association study by Canè *et al.* (2014). The same data were re-analyzed for a more accurate single nucleotide polymorphism (SNP)-based genome-wide association study (GWAS).

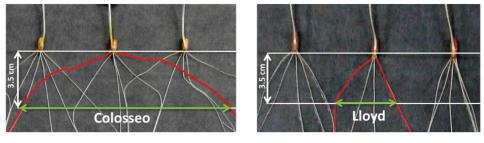
Field evaluation of RGA On 22 October 2014, the Unibo-DP was planted in Cadriano, Po Valley, Italy (44°33'N and 11°27'E) under a complete randomized block design with two replicates and optimal agronomic management. The soil is a highly fertile agricultural loam (Udic Ustochrepts, fine silty, mixed, mesic). Accessions were grown in 3-m-long twin rows (20 cm apart) with plants spaced apart by 10 cm, for a final density of 200 plants m⁻². Forty accessions were selected for root phenotyping in the field as follows: (i) two sets of 12 accessions each were chosen from the top- and bottom-ranking lists of the 183 accessions sorted for seminal RGA (Canè et al., 2014) and (ii) an additional 16 accessions used as parents of mapping populations developed at UNIBO, plus Colosseo, Lloyd, Meridiano, and Claudio parental lines were considered. The relationship between RGA at the seedling stage and the mature RGA phenotypes of field-grown plants at the post-flowering stage was investigated by using a 'shovelomics' approach (Trachsel et al., 2011). Root phenotyping in the field was carried out at the end of the flowering stage (Zadocks 69). In the selected plots, a 40 cm long and 30 cm wide central section was chosen for a 30 cm deep core-root system excavation. Root soil cores of 16-20 plants were water-soaked overnight, rinsed in clean water and allowed to dry. From each plot, the eight undamaged plants most homogeneous in tiller number along the row plot were imaged one by one (see Fig. 1B for an example). Images were processed with the Rest software (http://www.plant-imageanalysis.org/software/rest).

Molecular data and genetic map construction

The SSR and DArT[®] profiles of the Co×Ld and Mr×Cd RIL populations, their parents, and the Unibo-DP accessions (Maccaferri *et al.*, 2008, 2011; Mantovani *et al.*, 2008) were integrated with the high-density Infinium[®] iSelect[®] Illumina 90K SNP array (Wang *et al.*, 2014; Maccaferri *et al.*, 2015). DNA was extracted from a bulk of 25 one-week-old seedlings per accession using the DNeasy 96 Plant Kit (Qiagen GmbH, Hilden, Germany). The final array included 81 587 transcript-associated SNPs, 8000 of which are durum-specific SNPs (Wang *et al.*, 2014). Genotyping was

A) Root growth angle of seminal roots in 9-day-old seedlings

measured as the linear distance (in green) between the two most external roots (in red) of each plantlet at 3.5 cm from the seed tip and then converted to degrees.



B) Root growth angle measured on field-grown adult plants (nodal roots, REST software.)

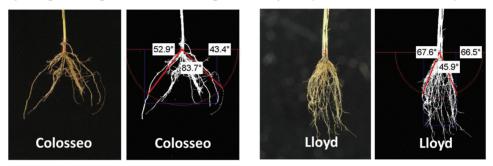


Fig. 1. Measurement of root growth angle (RGA) in seminal roots at the seedling stage and in excavated roots of field-grown plants at the end of flowering. The two parental genotypes Colosseo and Lloyd, highly contrasted for RGA at both stages, are shown as an example. Direct measurements were carried out for seminal root RGA, while software-aided digital measurements were carried out for adult root systems (REST software). (This figure is available in colour at *JXB* online.)

performed by TraitGenetics GmbH (Gaterlesleben, Germany). including duplicates of parental lines and reference cultivars. Briefly, the Infinium[®] iSelect[®] Illumina 90K SNP assay was performed on the 24 sample HD BeadChip format where genomic DNA is isothermally amplified, fragmented, and hybridized to the BeadChip. The amplified and fragmented DNA samples anneal to locus-specific, SNP-specific oligomers that are immobilized on the beads, then single-base extension at the SNP position is carried out on the BeadChip using the captured DNA as template and incorporating detectable and spectrally distinct fluorescent dyes associated with the two alternative alleles. The iScan® system laser-excites and laser-scans the BeadChip array at ultra-high resolution, thus providing the light-emitted raw quantitative output for each bead. The automated process is carried out according to the manufacturer's protocols (Illumina, Inc., San Diego, CA, USA). SNP clustering and genotype calling were performed using GenomeStudio v2011.1 software (Illumina, Inc.). Since the GenomeStudio v2011.1 was developed for diploids, a specific cluster file was developed by manually adjustment of the Illumina Genome Studio cluster calling for each marker to capture variation in data correctly. The cluster file is available upon request from Martin Ganal (TraitGenetics GmbH, Gatersleben, Germany). For the two mapping populations, linkage maps were based on the joint analysis of SSR, DArT[®], and SNP data from the Illumina 90K SNP assay that were assembled using a common mapping methodology described in Maccaferri et al. (2015). Briefly, the common mapping procedure involved the use of stringent thresholds for marker and line quality data filtering and the use of Carthagene v1.2.3 for gouping and mapping. The two genetic maps used for QTL analysis are reported as supplemental material in Maccaferri et al. (2015).

A tetraploid consensus map incorporating SSR, DArT[®], and SNP markers from 13 mapping populations (Maccaferri *et al.*, 2015) has been used to compare QTL results across mapping populations and the durum association panel and with those reported in bread wheat. The consensus map includes SSR, DArT[®], and SNP

markers for a total of 30 244 mapped markers and 2631 cM (11.5 markers cM^{-1} , on average).

Statistical analysis

ANOVA (general linear model, unbalanced ANOVA) was conducted for all RSA traits including blocks, replicates, and genotypes. Block effect, when significant, was accounted for by linear regression correction. The weight of the same seeds used in the RSA experiments (iTGW) was used as a covariate to correct for variation in seed weight in the root experiments. In addition, iTGW was also subjected to QTL analysis, similarly to the other root traits analysed.

ANOVA was carried out in Minitab[®]15, Minitab Ltd, Coventry, UK.

Heritability (h^2) was calculated on a mean basis across three replications according to the formula:

$$h^2 = \sigma_{\rm G}^2 \left(\sigma_{\rm G}^2 + \sigma_{\rm E}^2 / r \right)$$

where: σ_G^2 =genetic variance, σ_E^2 =residual variance, *r*=number of reps, σ_G^2 =(MS_{genotypes}–MS_{residual})/*r*, σ_E^2 =MS_{residual}, and MS stands for mean square value.

The coefficient of variation (CV) was defined as the SD (standard deviation)/mean: $CV=\sigma/\mu$; it assesses the extent of residual variability in relation to the phenotypic mean.

QTL analysis was carried out based on single marker analysis and multiple interval mapping (MIM; Kao *et al.*, 1999) in Windows QTL Cartographer v2.5 (http://statgen.ncsu.edu/qtlcart/WQTLCart.htm). Details on the MIM QTL searching model are given in Supplementary Text S1 at *JXB* online.

For each QTL, the effect was computed as the phenotypic difference between the mean values of the RIL groups homozygous for the two parental alleles at the QTL peak position. As a general rule, two QTLs identified for different root traits were considered as a single QTL cluster whenever their confidence intervals overlapped, QTL peaks mapped within a 15 cM interval, and the direction of the QTL effects showed consistency across traits.

In the Unibo-DP, a total of 19 815 bi-allelic SNP and DArT[®] markers were mapped to unique positions. Details on the GWAS model adopted are given in Supplementary Text S2. Briefly, three levels of significance were considered for reporting the GWAS QTLs: (i) marker-wise $P \le 0.01$ ($-\log_{10} P \ge 2.00$) for suggestive QTLs; (ii) marker-wise $P \le 0.001$ ($-\log_{10} P \ge 3.00$) for nominal QTLs; and (iii) experiment-wise $P \le 0.05$ ($-\log_{10} P \ge 4.00$) for major QTLs.

Due to the relatively high density of informative SNPs mapped from the panel (13 823 mapped SNPs with minor allele frequency ≥ 0.05 , corresponding to ~ 5.32 SNPs cM⁻¹) compared with the linkage disequilibrium (LD) decay rate observed in the panel (decay to $r^2=0.3$ at 2.20 cM distance, on average), the experiment-wise GWAS significance threshold was set according to the actual number of 'independent SNP tests'. The latter was estimated in Haploview using the tagger function using an r^2 tag threshold of 0.3 (Mackay, 1996). The total number of tag-SNPs was equal to 773, rounded to 1000, hence the experiment-wise, Bonferroni-corrected significance threshold at P=0.05 corresponded to a marker-wise threshold of $-\log_{10} P\ge 4.00$.

Since the observed density of mapped informative SNPs in the Unibo-DP (5.32 SNP/cM) far exceeded the minimum SNP density required to find QTL associations in the panel (based on the average genetic distance for LD decay to r^2 =0.3: 2.2 cM), GWAS QTLs were most frequently detected as multiple marker–trait associations of SNPs in LD with each other. For each QTL, the most associated SNP was considered as the QTL-tagging SNP marker.

The relative positions of RSA QTLs identified in our mapping populations, Unibo-DP, and in previous linkage studies in wheat (Laperche *et al.*, 2006; Kubo *et al.*, 2007; Sanguineti *et al.*, 2007; Guo *et al.*, 2012; Hamada *et al.*, 2012; Ren *et al.*, 2012; Bai *et al.*, 2013; Christopher *et al.*, 2013; Liu *et al.*, 2013; Cao *et al.*, 2014; Atkinson *et al.*, 2015; Petrarulo *et al.*, 2015) were compared based on the projected QTL peaks and confidence intervals on the tetraploid consensus map used as a common reference (Maccaferri *et al.*, 2015). Marker and QTL projection was carried out in Biomercator version 4.2, using QTLProj subroutine (BioMercator V4; Sosnoswski *et al.*, 2012).

Co-location of RSA QTLs with QTLs for grain weight and grain yield

The Unibo-DP was field-assessed in 15 field experiments carried out in the Mediterranean region in 2004 and 2005. The thousand grain weight (TGW) and grain yield (GY) data used were those published in Maccaferri *et al.* (2011). Trials were classified into three categories of (i) three low-yielding trials with yield from 0.9 t ha⁻¹ to 2.9 t ha⁻¹); (ii) five medium-yielding trials with yield from 3.5 t ha⁻¹ to 4.6 t ha⁻¹); and (iii) seven high-yielding trials with yield from 5.4 t ha⁻¹ to 6.7 t ha⁻¹. TGW and GY data have been reanalysed for GWAS based on the new SNPs. All GWAS QTLs are presented as a unique QTL density plot for each of the three trial categories using 1 cM as plot unit and the QTL significance intervals.

Results

Phenotypic variation in Colosseo×Lloyd and Meridiano×Claudio for RSA

Co×Ld and Mr×Cl RIL populations were assessed for seminal RSA traits at the seedling stage. A summary of the phenotypic values for root and shoot traits is reported in Table 2 and distribution frequencies are reported as histograms in Supplementary Fig. S1. In both populations, the Anova (results reported in Table 2) detected highly significant ($P \le 0.01$) differences among the RILs for all considered traits. CVs for RSA traits ranged from 5 to 13% for most recorded traits [except for LRN (lateral root number per primary root) in Mr×Cl, with a CV of 18.5%], while heritability values (h^2) were mostly between 0.62 and 0.93, with only a few traits showing h^2 values <0.50 (Table 2). CVs were consistently lower and h^2 values higher in Co×Ld as compared with Mr×Cl.

Both RIL populations showed an RSA trait distribution approaching normality and strong transgressive segregation for most traits, with the only exception for the presence of the Rt6 that showed a distribution close to bi-modality (Supplementary Fig. S1). The fold range variation in trait values among RILs varied from 0.50 to 1.20 for most traits, with maximum fold range values for Rt6 (3.63) and LRN (2.40) in both populations, followed by total root length (TRL), average root length (ARL), and SDW in Co×Ld (Table 2; Supplementary Fig. S1).

RILs showed heritabilities of medium to high values (≥ 0.60) for most traits, with the highest values observed for RGA (h^2 =0.90 and 0.65 in Co×Ld and Mr×Cl, respectively), LRN (0.93 and 0.62, respectively), and SDW (0.91 and 0.70, respectively). Colosseo and Lloyd differed markedly for RGA, SL, SDW, primary and total root length (PRL and TRL, respectively), and presence/absence of Rt6 (Table 1).

Among the four parents, Colosseo showed the widest RGA (40.9% wider than Lloyd, the parent with the narrowest RGA), as depicted in Fig. 1 which provides a schematic representation of the methodologies used to evaluate RGA, and in Fig. 2, which provides a comparison between RSA of Colosseo and Lloyd at both seedling and adult plant stages. As compared with Colosseo, Lloyd was characterized by significantly longer coleoptiles (SL, +29.1%) and primary roots (PRL, +32.7%), more lateral root tips (LRN, +124%), and higher SDW (42.0%, respectively), while RDW was higher in Colosseo (+49.2%). This resulted in a higher root to shoot ratio in Colosseo than in Lloyd (1.22 versus 0.64). The majority of Colosseo seedlings (70.0%) showed the presence of Rt6, while Lloyd seedlings did not develop this root. As compared with Colosseo and Lloyd, Meridiano and Claudio differed less for RSA traits, showing significant differences for TRN, TRL, ARL, LRN, and RGA, but not for PRL, Rt6, RDW, RSR, SL, and SDW (Table 2, Fig. 2).

The phenotypic values for RSA traits observed in the two mapping populations were in the range of those observed for the Unibo-DP as reported in Canè *et al.* (2014). Notably, for RGA, the panel showed a wider range of values than those observed in the two RIL populations (48–147 ° versus 47–110 ° in Co×Ld and 68–129 ° in Mr×Cd), as depicted in Fig. 2.

The relationship between RGA evaluated at the seedling stage and RGA at the adult stage measured under field conditions (nodal roots) was assessed on selected contrasting 'tails' from the RGA phenotypic distribution of the Unibo-DP accessions and in a small panel of parents of mapping populations. Summary statistics of comparison between RGA of seminal roots at the seedling stage and of the adult root system **Table 1.** Summary of acronyms used for the traits measured inthis study

Acronym	Trait	Measuring unit
Seedling traits		
ARL	Average root length	cm
iTGW	Thousand grain weight of the seed used	g
	in the root seedling evaluation experiment	
LRN	Lateral root number per primary root	n
PRL	Primary root length	cm
PRD, TRD, ARD	Primary, total, average root diameter	mm
PRS, TRS, ARS	Primary, total, average root surface	mm ²
PRV, TRV, ARV	Primary, total, average root volume	mm ³
RDW	Root dry weight	mg per plant
RGA	Root growth angle	0
RSR	Root to shoot ratio	ratio
Rt6	Presence of the sixth asymmetric	% of seedlings
	seminal root	
SDW	Shoot dry weight	mg per plant
SL	Shoot length	cm
TRL	Total root length	cm
TRN	Total root number	n
Field traits		
TGW	Thousand grain weight from the field	g
	experiments	
GY	Grain yield	t ha ⁻¹
Other acronyms		
GWAS	Genome-wide association mapping	
RSA	Root system architecture	
RGA	Root growth angle	
NUE	Nitrogen use efficiency	
SNP	Single nucleotide polymorphism	
UNIBO-DP	UNIBO-Durum Panel	
WUE	Water use efficiency	

from the field are reported in Table 3. On average, the RGA of adult root systems under field conditions is narrower than the RGA on seminal roots on screening plates. The group of accessions selected for narrow seedling RGA showed a mean field RGA equal to 51 ° versus a mean seedling RGA of 73 ° (with a 29.7% reduction for field RGA) while the group selected for wide seedling RGA had a mean field RGA equal to 64 ° versus a mean seedling RGA of 125 ° (48.7% reduction for field RGA). The parents of mapping populations showed a mean field RGA equal to 58 ° and a mean seedling RGA equal to 106 ° (44.9% reduction for field RGA). The regression of adult RGA on seedling RGA was highly significant, with an R^2 value equal to 21.8% (Fig. 3). Correlations between RSA traits are reported in Supplementary Table S1 and commented on in Supplementary Text S3.

QTLs for root and shoot traits

Two high-density linkage maps were produced for the Co×Ld and Mr×Cd populations, including 7946 and 5970 mapped markers, respectively. Total map length was 2064 cM for Co×Ld and 2248 cM for Mr×Cd. Each map covered \sim 70% of the durum wheat genome relative to the reference consensus map (Maccaferri *et al.*, 2015), hence revealing

extensive identity by descent regions. In the Unibo-DP, a total of 13 196 polymorphic SNPs showed a minimum allele frequency (MAF) ≥ 0.05 and were uniquely mapped in the consensus map.

For all traits, both RIL populations showed numerous QTLs (Supplementary Tables S2, S3). The Co×Ld population showed 48 QTLs (20 suggestive and 28 nominal QTLs) which were grouped into 26 QTL clusters. The Mr×Cd population showed 81 QTLs (49 suggestive and 32 nominal QTLs), for a total of 44 QTL clusters.

The detected QTLs showed a broad range of R^2 values (from 4.0 to 26.9%, Supplementary Tables S2, S3). Determination of the distribution of QTLs between the two genomes showed a relative enrichment for the B genome, with 28 and 42 QTL clusters found in the A and B genome, respectively.

The association mapping panel (not phenotyped for LRN, nor root diameter, surface, and volume) showed 201 QTLs, five of which reached the experiment-wise threshold (-log $P \ge 4$), 26 exceeded the marker-wise threshold of $-\log P \ge 3$, and 170 had a marker-wise -log P-value of between 2 and 3 (suggestive QTLs). The GWAS QTLs from the first two categories were considered as nominal QTLs, while the remaining 170 were classified as suggestive QTLs. All the GWAS QTLs were then grouped into 112 QTL clusters. Considering RSA traits as a whole, the high-density SNP genotyping allowed us to detect many additional QTL clusters as compared with (112 versus 48) those reported in Canè et al. (2014) using SSR and DArT[®]. However, the relative QTL mapping efficiency of SNP versus SSR and DArT[®] varied depending on the trait: for RGA, a highly heritable trait, 76% of the QTLs detected with the high-density SNP map were also detected by Canè et al. (2014). Details of GWAS QTL results for the RSA traits are reported in Supplementary Table S4.

All the QTL information was cross-referenced by projecting the QTL confidence intervals on the tetraploid wheat consensus map (Maccaferri *et al.*, 2015). Other RSA QTLs from published studies in wheat were also cross-referenced, whenever possible, by projecting common markers onto the tetraploid wheat consensus. Chromosome arms with a high number of RSA QTLs were 1AS, 1BS, 2A centromeric, 2BS, 2BL, 3A centromeric, 3AL, 3BS, 5AS, 5BS, 6AS, 6BS, 6BL, 7AS, and 7BL (Supplementary Fig. S2).

Main QTL clusters for root length and number

The prioritization of RSA QTL clusters relied on the significance level of single-component QTLs, the consistency of their effects on root traits across durum and bread wheat, as well as the presence of significant overlap with field TGW and GY from data of the 15 Mediterranean environments available for the Unibo-DP (Maccaferri *et al.*, 2011). Due to the genetic control of RGA which is largely independent from that of the other RSA traits, QTLs for RGA have been prioritized separately.

Among the 85 RSA QTL clusters for root vigour (root length, weight, and/or number) identified in this study, 20 appeared of particular interest based on the cumulated proof for the presence of RSA QTLs across the genetic materials

Table 2. Summary statistics for the root and shoot traits measured at the seedling stage in the parents and recombinant inbred lines (*RILs*) of two durum wheat populations

Genotype	TGW ^a	TRN	Rt6	PRL	TRL	ARL	LRN	RGA	SL	SDW	RDW	RSR
	(g)	(n)	(%)	(cm)	(cm)	(cm)	(n)	(°)	(cm)	(mg per plant)	(mg per plant)	(ratio)
Colosseo×Lloyd												
Colosseo	53.5	5.6	70.0	20.5	118.2	18.1	6.9	107	11.4	13.8	16.9	1.22
Lloyd	51.3	4.8	0.0	27.2	130.8	21.2	15.5	76	14.6	18.6	12.0	0.64
RILs												
Mean	50.9	5.30	36.3	22.9	96.9	18.6	9.5	82	13.8	14.0	13.3	0.96
Minimum	40.3	4.22	0.0	14.9	52.1	10.5	3.3	47	10.7	7.6	7.6	0.72
Maximum	64.0	6.09	100.0	33.1	143.2	28.2	19.7	109	16.2	20.3	18.2	1.33
Fold range	0.47	0.35	2.75	0.79	0.94	0.95	1.73	0.76	0.40	0.91	0.80	0.64
Significance ^b	**	**	**	**	**	**	**	**	**	**	**	**
CV (%)	5.6	4.9	5.40	8.62	8.95	8.09	12.9	6.3	4.4	8.6	7.40	10.0
h^2	0.87	0.73	0.88	0.48	0.63	0.62	0.93	0.90	0.84	0.82	0.76	0.66
Meridiano×Claud	io											
Meridiano	51.6	5.6	30.2	32.9	133.9	22.5	2.2	85	13.7	13.5	12.3	0.91
Claudio	54.0	4.9	28.5	32.7	117.4	26.5	15.0	104	13.8	12.4	13.3	1.07
RILs												
Mean	48.3	5.0	16.0	30.7	122.5	22.9	10.4	98	14.7	12.2	11.6	0.96
Minimum	39.8	4.4	0.0	24.7	95.8	17.7	2.2	68	10.9	9.6	8.6	0.70
Maximum	63.4	5.7	58.1	34.5	142.4	26.2	27.2	129	15.6	14.7	13.7	1.22
Fold range	0.49	0.26	3.63	0.32	0.38	0.37	2.40	0.61	0.32	0.42	0.44	0.54
Significance ^b	**	**	**	**	**	**	**	**	**	**	*	**
CV (%)	6.4	13.4	13.5	5.7	7.12	6.6	18.5	13.5	7.0	12.6	13.8	13.0
h^2	0.79	0.62	0.64	0.52	0.56	0.32	0.62	0.65	0.64	0.70	0.62	0.56

^a Root trait acronyms: TGW, thousand grain weight; TRN, total root number; Rt6, presence of the sixth root; TRL, total root length; ARL, average root length; LRN, lateral root number on the seminal primary root; RGA, root growth angle; SL, shoot length; SDW, shoot dry weight; RDW, root dry weight; RSR, root to shoot ratio.

^b Significance of differences among RIL lines: $*P \le 0.05$; $**P \le 0.01$.

that were explored and co-location with GY QTLs in droughtstressed environments. Features of these QTL clusters are summarized in Table 2 and detailed in Supplementary Fig. S2. Examples of QTL cluster maps for chromosomes 2B, 4B, 6A, and 7A are provided in Fig. 4. Twelve clusters showed consistent effects on both root length and number (RSA_ QTL_clusters_2#, 3#, 4#, 6#, 7#, 8#, 11#, 12#, 13#, 14#, 17#, and 18#). Additionally, RSA_QTL_clusters_1#, 5#, 9#, 10#, 15#, 16#, 19#, and 20# were considered specific for root length/elongation.

QTLs for root growth angle

Biparental mapping identified 12 RGA QTLs, six in Co×Ld and six in Mr×Cd. Details of RGA QTL features are reported in Table 5. As expected based on the four parental phenotypes, the narrow-angle phenotype was mostly contributed by Lloyd. Three QTLs were classified as major based on additive effects on root angle ≥10 ° and LOD score >3: *QRga.ubo-2B* and *QRga.ubo-6A* in Co×Ld (R^2 =11.7% and 17.8%, respectively) and *QRga.ubo-4B* in Mr×Cd (R^2 =13.8%, see Table 5).

As compared with the mapping populations, GWAS on the Unibo-DP revealed 17 novel RGA QTLs, nine of which showed an R^2 coefficient >5.0% (Table 5). Therefore, limited overlap was observed between the QTLs identified in biparental and association mapping. The only QTL that showed a

tight co-location between one mapping population (Co×Ld) and the association panel was QRga.ubo-6A.2. However, overlap between RGA QTLs in tetraploid wheat and previously reported QTLs in bread wheat was observed on chromosomes 2A, 2B (two cases), 3B, 6B (two cases), and 7A.

Details of RGA phenotypic data and RGA QTL-tagging SNP allelic distribution in the Unibo-DP germplasm subgroups are reported in Supplementary Tables S5 and S6. A schematic representation of contrasting shallow and deep wheat root ideotypes with the QTLs for seminal RGA identified in this study which can potentially contribute (through marker-assisted selection) is provided in Supplementary Fig. S4. The SNP alleles for narrow RGA were present at high frequency (≥ 0.5) for nine QTLs, at balanced frequency (0.20–0.50) for seven QTLs, and at relatively low frequency (0.10-0.20) for four QTLs only. This indicates that the alleles contributing a narrow RGA are already present at relatively high frequencies in the cultivated accessions. Here, it is also worth noting that QTL effect and significance inferences in association mapping are both influenced by allele frequency in populations, particularly as to the precision of allelic variant effect estimates and the robustness of association. Considering *QRga.ubo-6A.2*, which is one of the most important RGA QTLs found both in Col×Ld and Unibo-DP at high R^2 levels, it can be noticed that the narrow-angle allele was absent in the ICARDA germplasm for dryland areas

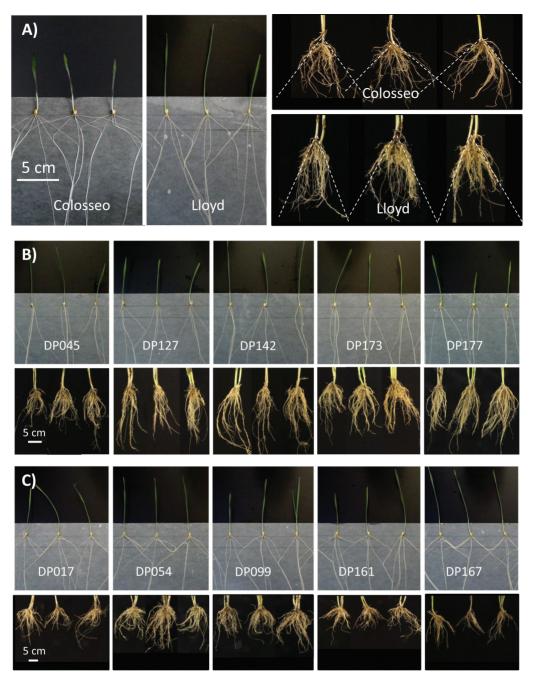


Fig. 2. Root system architecture (RSA) of seedlings and field-grown plants of durum wheat parental lines and accessions from the elite Unibo-DP showing contrasting phenotypes for RSA features. (A) Comparison of RSA for the parental lines Colosseo (left) and Lloyd (right) at the adult stage in the field. Dotted lines originating at the crowns were traced to delimit 95% of the 2-D projected area of the root system (as suggested in the REST software manual). (B) Comparison of seedling and adult RGA for the five Unibo-DP accessions with the narrowest RGA according to seedling measurements. (C) Comparison of seedling and field RGA for the five Unibo-DP accessions with the widest RGA according to seedling measurements. (This figure is available in colour at *JXB* online.)

(Syrian germplasm, directly derived from the West Asian native germplasm) while being present at higher than average frequencies in the modern high-yielding germplasm including the most recent germplasm from CIMMYT/ICARDA programmes and the Italian and DesertDurum[®] cultivars. A similar allelic distribution was observed for the second most relevant QTL found in the association panel, *QRga.ubo-2A* (Supplementary Table S6).

Additionally, interesting QTLs with sizeable effects ($\geq 10^{\circ}$) on RGA were identified by GWAS on chromosomes 1B, 2A,

and 7A. For these QTLs, the narrow-angle allele was present at low frequency in the elite germplasm.

Relationships between QTLs for RSA, GY, and TGW

Co-locations between RSA QTLs and QTLs for TGW and GY were assessed based on the GWAS reanalysis of field data from the evaluation of the Unibo-DP in 15 field trials under a wide range of water availability and grain yield (from 0.99 t ha^{-1} to 6.74 t ha^{-1}) in the Mediterranean region (for more details, see

Maccaferri *et al.*, 2011). In our study, the addition of 13 196 SNPs to the 957 SSR and DArT markers used by Canè *et al.* (2014) greatly improved detection of QTLs for GY, a trait with medium to low heritability (h^2 from 0.42 to 0.67) and TGW.

Among the 20 RSA QTL clusters identified in this study, only two had no impact on either field GY or TGW

Table 3. Summary statistics for the root growth angle (RGA)at seedling and adult plant stages in 24 selected accessions ofthe Unibo-DP and in 16 accessions used as parents of mappingpopulations

Accessions	Seedling (°)	Field ^a (°)
Narrow RGA (12 accessions) ^b		
Mean	73.1	51.4
Minimum	66.3	35.3
Maximum	78.3	62.8
CV (%)	10.4	17.7
Wide RGA (12 accessions)		
Mean	125.3	64.2
Minimum	115.0	49.3
Maximum	139.2	76.2
CV (%)	12.4	15.9
Parental lines (16 accessions)		
Mean	106.1	58.4
Minimum	82.5	42.3
Maximum	120.7	69.2
CV (%)	12.0	17.4
Colosseo	107.1	67.1
Lloyd	76.0	53.9
Meridiano	85.2	60.8
Claudio	104.4	58.5

^a Field RGA was evaluated at the end of flowering stage (Zadok 69) in a field trial carried out in a high-fertility alluvial loam soil (Cadriano, Italy).

^b Summary statistics are reported for three groups of accessions: (i) 12 bottom-ranking accessions from the 183 Unibo-DP accessions sorted for seminal RGA (Canè *et al.*, 2014); (ii) 12 top-ranking accessions from the 183 Unibo-DP accessions sorted for seminal RGA (Canè *et al.*, 2014); and (iii) 20 parents of RIL populations being developed at UNIBO, including Colosseo, Lloyd, Meridiano, and Claudio parents. (QTL_cls. 14# on chromosome 5A and 20# on chromosome 7B), while five were associated with peaks in GY but not final TGW (QTL_cls. 1#, 11#, 15#, 18#, and 19#) and six clusters co-located only with TGW (QTL_cls. 5#, 8#, 9#, 10#, 13#, and 17#). Finally, six QTLs co-located with both GY and TGW QTLs (QTL_cls. 3#, 4#, 6#, 7#, 12#, and 16#). Supplementary Fig. S3 reports the detailed co-location between RSA, GY, and TGW QTLs.

Among the 30 RGA QTLs, nine (*QRGA.ubo-2A.1*, *QRGA. ubo-2A.2*, *QRGA.ubo-2A.3*, *QRGA.ubo-2B.2/2B.3*, *QRGA. ubo-6A.2*, *QRGA.ubo-7A.1*, and *QRGA.ubo-7A.2/7A.3*, with the 2B.2/2B.3 and 7A.2/7A.3 QTLs to be considered as different QTLs with very close confidence intervals) mapped within relatively small intervals (≤ 10 cM) that co-localized with strong QTL peaks for GY and TGW, including GY in 3–4 environments and TGW in 4–7 environments, respectively (Supplementary Fig. S3). For *QRGA.ubo-2A.1*, *QRGA. ubo-2B.2/2B.3*, and *QRGA.ubo-7A.1*, it is worth noting that the co-locations of RGA QTLs with those for TGW and GY were specific for RGA.

For each RGA QTL with an R^2 value $\ge 5.0\%$ among those identified in the Unibo-DP. Supplementary Table S7 reports the concomitant effects on GY and TGW, based on the allelic variants present at the QTL-tag SNPs. Notably, the QTL alleles for wider root angle had a prevailingly negative effect on GY and/or TGW in the less favourable environments. The negative RGA versus GY association was observed for 35 (i.e. 49%) of the 72 QTL-environment pairs in eight drought- and heat-stressed environments (from Granada-2005, rainfed, to Tel Adja-2005, rainfed, for a total of eight environments with mean GY from 0.99 t ha^{-1} to 4.63 t ha^{-1}). Conversely, the same negative association was observed in only 17 (i.e. 27%) of the 63 QTL-environments pairs in seven medium to highly productive environments (from Rayack-2005, rainfed, to Kef-2005, irrigated, with mean GY from 5.61 t ha^{-1} to 6.78 t ha⁻¹). Moreover, the RGA versus GY association was consistently negative across the majority of the RGA QTLs in the most stressed environment (Granada-2005, rainfed) and for another environment with low GY (Tel-Adja-2005, rainfed). Importantly, ORga.ubo-7B, ORga.ubo-6A.1, and

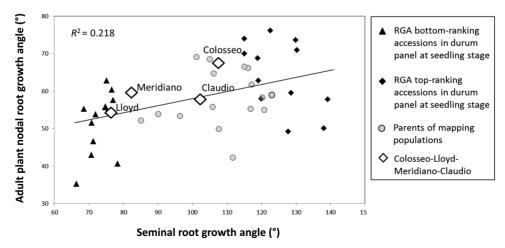


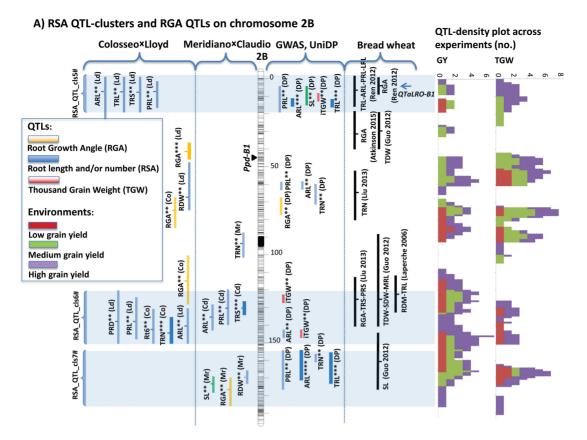
Fig. 3. Relationship between seminal root growth angle (RGA) at the seedling stage and adventitious root system growth in the adult plant under field conditions for a selection of 44 Unibo-DP accessions. Three accession groups were considered: (i) 12 Unibo-DP accessions bottom-ranking for seminal RGA; (ii) 12 Unibo-DP accessions top-ranking for seminal RGA; and (iii) 16 parents of mapping populations developed at UNIBO plus the parental lines of the Colosseo×Lloyd and Meridiano×Claudio RIL populations.

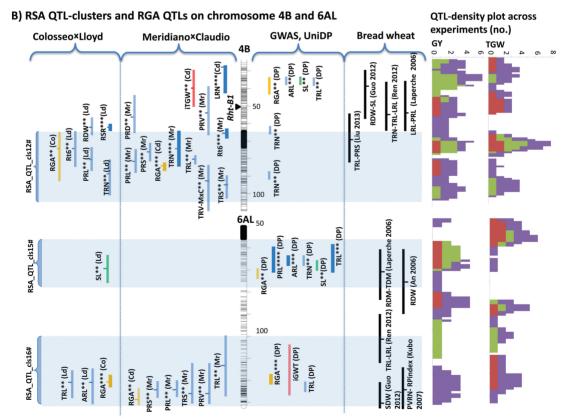
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Seminal root growth angle (RGA) QTLs are not included in the table.

RSA QTL cluster	Chr.	Interval ^a (cM)	Main RSA trait ^b	Colosseo×Lloyd ^c (trait)	Meridiano×Claudio (trait)	Unibo-DP (trait)	Wheat QTLs (trait, reference) ^d
RSA_QTL_cluster_1# ^e	1AS	0-20	RL	. 1	PRV, TRV	TRN, TGW	TRL, TRS, TRV, LRN (Tt1)
RSA_QTL_cluster_2#	1BS	0-25	RL, RN	TRL, TRV	TRS, TRV	ARL, SL, TRL, TRN	TRL, TRS, TRV, LRN (Tt1), TRL
							(Ta10)
RSA_QTL_cluster_3#	1BS	30-50	RL, RN	I	ARL, TRL	TGW	LRN (Tt1), TRN (Ta3), TRL (Ta6),
							TDW, SDW, RDW, TRN (Ta7), RDW
							(Tt11)
RSA_Q7L_cluster_4#	2AL	190-210	RL, RN, RGA	Ι	TRV, LRN, TGW	ARL, TRL, TRN, PRL, TGW, RGA	I
RSA_QTL_cluster_5#	2BS	0-20	RL	ARL, PRL, TRL, TRS	1	ARL, SL, PRL, TRL, TGW	TRL, PRL, LRL, qTaLRO-B1 (Ta2-9)
RSA_QTL_cluster_6#	2BL	120-150	RL, RN	ARL, PRD, PRL, Rt6, TRN	ARL, PRL, TRS	TGW	RGA, TRS, PRS (Ta6), TDW, SDW,
							MRL (Ta7), RDM, TRL (Ta10)
RSA_QTL_cluster_7#	2BL	155-185	RL, RN, RGA	Ι	RDW, SL, RGA	ARL, PRL, TRL, TRN	SL (Ta7)
RSA_QTL_cluster_8#	3AS	30-40	RL, RN	1	SL	ARL, PRL, TRL, TRN	1
RSA_QTL_cluster_9#	3AL	100-135	RL	I	I	ARL, PRL, TRL, TGW	TRL, TRS, TRV (Tt1), LRL, LRV,
							LRS (Ta5), MRL (Ta7), PRL (Ta9)
RSA_QTL_cluster_10#	3BS	0-10	RL	I	PRS	ARL, PRL, SL, TRL, TGW	1
RSA_QTL_cluster_11#	3BL	70-100	RL, RN	Ι	ARL, TRL	1	TRL, TRV, LRV, LRS (Ta5), TRL,
							PRL, LRV, TRS, TRN, RGA (Ta6)
RSA_QTL_cluster_12#	4BL	65-105	RL, RN	Rt6, PRL, TRN, RGA	Rt6, PRL, TRL, TRV, TRN, PRS, RGA	TRN	TRL, PRS (Ta6)
RSA_QTL_cluster_13#	5AS	25-65	RL, RN	Rt6, PRL, TRN	I	1	TRV, TRS, SDW, TGW (Ta5), TRN
							(Ta6), TRN (Ta7), TRN (Ta8)
RSA_QTL_cluster_14#	5AL	100-140	RL, RN	Ι	LRN, SDW	ARL, PRL, TRL	I
RSA_QTL_cluster_15#	6AL	55-75	RL	Ι	Ι	ARL, PRL, SL, TRL, TRN	RDM/TDM (Ta10)
RSA_QTL_cluster_16#	6AL	100-130	RL	ARL, TRL	PRL, TRL, PRS, TRS, PRV	TRL	SDW (Ta7), PVRN, RPindex (Tt11)
RSA_QTL_cluster_17#	6BL	75-100	RL, RN, SL	ARL, TRD, SL	ARL, PRL, PRS, ARL, Rt6	TRL	MRL (Ta7)
RSA_QTL_cluster_18#	7AL	100-120	RL, RN	Ι	PRD, PRV, TRS	PRL	TRN (TaG)
RSA_QTL_cluster_19#	7BL	155-175	RL	I	PRS, PRL	PRL, RGA	TRL (Ta6), RDW, SDW, TDW, MRL
							(Ta7)
RSA_QTL_cluster_20#	7BL	185–220	RL	TRL, TRS, ARL	I	I	TDW, RDW (Ta7), LRN/PRL (Ta10)

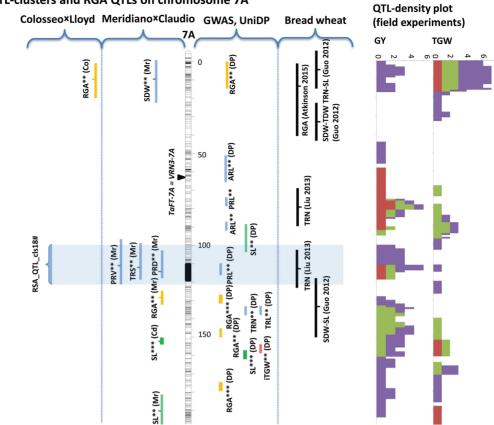
^a Chromosome interval refers to the tetraploid wheat consensus map (Maccaferri *et al.*, 2015).
^b Main RSA trait acronyms: RL, root length; RN, root number; RGA, root growth angle.
^c Root trait acronyms: TGW, thousand grain weight; TRN, total root number; Rt6, presence of the sixth root; TRL, total root length; ARL, average root length; LRN, lateral root number on the central primary root; RGA, root growth angle.
^c Root trait acronyms: TGW, thousand grain weight; TRN, total root number; Rt6, presence of the sixth root; TRL, total root length; ARL, average root length; LRN, lateral root number on the central primary root; RGA, root growth angle; SL, shoot length; SDW, shoot dry weight; RSR, root to shoot ratio.
^d Literature references are listed in full in Supplementary Fig. S2.
^e The QTL cluster acronyms have been differentiated from the nomenclature currently used for single-trait QTLs by using consecutive numbering followed by the # symbol.





QRga.ubo-7A.4 showed a consistently negative RGA versus GY association of effect across the majority of environments, including the medium- to high-yielding ones, while the QTLs *QRga.ubo-7A.5*, *QRga.ubo-2B.2*, and *QRga.ubo-3A* showed a

negative RGA versus GY association mostly in highly stressed environments. The relationship between RGA and TGW was not as clear as that with GY, possibly due to the highly compensating interaction between yield components in wheat.



C) RSA QTL-clusters and RGA QTLs on chromosome 7A

Fig. 4. Genetic maps of root system architecture (RSA) QTL clusters and root growth angle (RGA) QTLs for chromosomes 2B, 4B, 6A, and 7A. The reference map is the tetraploid consensus map reported by Maccaferri *et al.* (2015). RSA and RGA QTLs from Co×Ld and Mr×Cl RIL populations, GWAS QTLs from Unibo-DP. and previously published QTLs from bread wheat studies have been projected onto the reference maps. Single-component QTLs are reported as vertical bars corresponding to confidence intervals. RSA QTL clusters are highlighted by horizontal shaded banding. GY and TGW QTLs in the trait acronyms are the same as in Table 1. QTL significance levels are highlighted using ** for suggestive QTLs, *** for nominal QTLs, and **** for GWAS experiment-wise significant QTLs (Unibo-DP only).

Discussion

Our study reports novel QTLs for RSA traits at the seedling stage in tetraploid wheat, provides a comprehensive survey of the QTLome for RSA features in wheat, and demonstrates that RSA phenotypes such as root length and RGA are governed by allelic variation at major and minor QTLs.

From a methodological standpoint, phenotyping seminal roots proved to be highly accurate and cost-effective for evaluating hundreds of lines (540 in total) as required in QTL discovery studies. Previous studies have investigated RSA on seedlings in wheat and other crops and have shown that QTLs for root features at this stage can be predictive of RSA of field-grown plants (Tuberosa et al., 2002; Landi et al., 2010; Mace et al., 2012; Li et al., 2015; Richard et al., 2015; Uga et al., 2015). Notably, in wheat and barley it has been demonstrated that the functional importance of the seminal roots as compared with the nodal roots increases under drought and other environmental stress (Belford, 1987; Manske and Vlek, 2002; Manschadi et al., 2010). An actively growing seminal root apparatus is important for early vigour and crop establishment in dryland areas (Lopez-Castaneda et al., 1995; Reynolds and Tuberosa, 2008). The seminal root system also plays an important role under high-density planting in high-input conditions (Manske and Vlek, 2002).

The main core of the results and QTLs reported herein refer to phenotypic characterization and QTL identification carried out on wheat seminal roots at an early growth stage (9-day-old seedlings grown in water). The objective here is to identify OTLs with strong and constitutive expression, in order to be able to translate their effects to wheat plants grown under various conditions, including water- and nutrient-stressed fields. Heritability in experiments carried out under standard controlled conditions is higher as compared with experiments where the environmental component of variation is more sizeable, such as in the case of seedlings grown in soil and/or at contrasting temperatures. Environment-specific, adaptive QTLs can be missed by analysis under standard conditions. However, this is counterbalanced by gains in reproducibility and the capability to assess effectively the genetic value of the trait, therefore increasing the capacity to detect the underlying constitutive QTLs (Collins et al. 2008).

GWAS in cultivated durum wheat germplasm

In wheat, QTL detection and genomics-assisted applications have recently benefited from novel marker platforms (Tuberosa and Pozniak, 2014) such as the Illumina 90K wheat

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	Chr.	Cl ^a	QTL-tag SNP	CoxLd ^b	SNP°	Mr×Cl ^d	SNP®	Association panel/	SNP ^g	Wheat QTLs ⁿ
		(cM)	(ID)	(effect, <i>R</i> ² , isignif.i) [/]	(allele)	(effect, <i>R</i> ² , signif.)	(allele)	(effect, <i>R</i> ² , signif.)	(allele)	(reference)
QRga.ubo-1B	đ	61-62-68	IWA7700	NS	C/T	NS	T/T	+7.2 (4.2) **	C/T	
				I		I				
				(1	1			ļ	
QRga.ubo-2A.1	2A	7-10-12	IWB16988	NS	1/1	NS	1/T	+7.7 (4.7) **	C/T	
				I		I				
QRga.ubo-2A.2	2A	97–99–102	IWA1597	NS	T/T	NS	T/T	+10.8 (3.9) **	C/T	(Ta6)
QRga.ubo-2A.3	2A	208-211-213	IWB20424	NS	AA	NS	AA	+9.5 (7.5) ***	A/G	(Ta3)
QRga.ubo-2B.1	2B	37-43-45	IWB4604	-10.0 (11.7) ***	T/C	NS	T/C	NS	C/T	
QRga.ubo-2B.2	2B	70-72-75	IWB73449	NS	A/A	NS	G/G	+8.2 (6.2) **	A/G	
QRga.ubo-2B.3	2B	68-77-87	IWB39220	+6.0 (6.2) **	A/A	NS	G/G	NS	A/G	
QRga.ubo-2B.4	2B	100-114-130	IWB62718	+5.0 (4.4) **	G/A	NS	A/A	NS	A/G	(Ta6)
QRga.ubo-2B.5	2B	175-176-183	IWB13830 = IWB7663	NS	AG	+5.8 (6.7) ***	A/G	+4.70 (2.4) *	A/G	
QRga.ubo-3A	ЗA	171-173-176	IWB70610	NS	A/A	NS	A/A	-6.9 (5.5) **	A/G	
QRga.ubo-3B	3B	181-188-196	wPt-5947	NS	1/0	-4.6 (4.2) **	1/0	NS	I	
QRga.ubo-4A	4A	154-156-159	wPt-0538	NS	1/1	NS	1/1	-12.5 (4.2) **	1/0	
QRga.ubo-4B.1	4B	33-35-38	IWB73831	NS	T/C	NS	T/C	-6.7 (4.6) **	C/T	
QRga.ubo-4B.2	4B	44-47-49	IW/B69501	NS	T/C	NS	C/C	-8.6 (3.8) **	C/T	
QRga.ubo-4B.3	4B	66-75-91	IW/B60481	+4.2 (2.4) **	C/T	NS	C/C	NS	C/T	
QRga.ubo-4B.4	4B	82-82-86	IWB12276	NS	A/A	-8.2 (13.8) ***	A/G	NS	A/G	
QRga.ubo-5B.1	5B	2-3-13	IWB75279	NS	1/1	-5.6 (6.1) ***	0/1	NS	1/0	
QRga.ubo-5B.2	5B	47-49-52	IWB24957	NS	C/C	NS	C/C	+15.0 (4.6) **	C/T	
QRga.ubo-6A.1	6A	70-72-75	IWB50963	NS	A/G	NS	A/G	-11.9 (5.2) **	A/G	
QRga.ubo-6A.2	6A	119-122-123	IWB71119	+10.0 (17.8) ***	G/A	NS	G/A	-10.6 (8.8) ***	A/G	
QRga.ubo-6A.3	6A	127-130-136	IWB24237	NS	T/T	-5.8 (7.0) **	T/C	NS	C/T	
QRga.ubo-6B.1	6B	20-22-25	IWB62692	NS	A/A	NS	A/A	+6.8 (4.8) **	A/G	(Ta3)
QRga.ubo-6B.2	6B	143-145-147	IWB 13062	NS	G/G	NS	G/G	-7.0 (4.5) **	A/G	(Ta3)
QRga.ubo-7A.1	7A	4-10-21	wPt-9207	+6.0 (5.5) ***	0/1	NS	1/1	NS	1/0	(Ta4)
QRga.ubo-7A.2	7A	128-130-133	IWA2752	NS	C/C	NS	C/C	+11.8 (7.1) ***	C/T	
QRga.ubo-7A.3	7A	125-131-135	IWB35428	NS	T/T	+5.6 (6.5) **	C/-	NS	C/T	
QRga.ubo-7A.4	7A	145-147-150	IWB56095	NS	C/T	NS	-/C	-7.2 (5.3) **	C/T	
QRga.ubo-7A.5	7A	177-180-183	IWB3169	NS	C/C	NS	T/C	+10.4 (7.2) ***	C/T	
QRga.ubo-7B	7B	175-177-180	IWB371	NS	G/G	NS	G/G	+12.8 (5.1) **	A/G	

non-significant. ^e SNP alleles in the mapping population: Meridiano allele (first)/Claudio allele (second). ^e SNP allelic effect (*a*) reported as double-dose (*2a*) substitution of the A or C allele (reported first) versus the G or T allele (reported second). ^g SNP alleles in Unibo-DP. The allele with the highest frequency is reported as bold italic font. ^h Literature references are listed in Supplementary Fig. S2. ^h RGA is expressed as degress while *R*² is expressed as a percentage; QTL effect significance; RIL populations, **LOD 2-3; ***LOD>3; GWAS in Unibo-DP, **marker-wise *P*≤0.01; ***marker-wise *P*≤0.001). SNP alleles in the RiL population: Colosseo allele (first)/Lloyd allele (second). SNP allelic effect (a) reported as double-dose substitution (2a) of the Meridiano versus Claudio allele. A positive sign indicates that the Meridiano allele increases the trait and vice versa; NS,

SNP array, that allows for highly accurate genome-wide scans at 0.2 cM per marker (Wang *et al.* 2014; Maccaferri *et al.*, 2015), hence greatly facilitating and enhancing the accuracy of comparative analysis and QTL cross-referencing.

The analysis of intermarker LD (the genetic association between markers specifically estimated in the Unibo-DP genotypes) based on the high-density SNP map confirmed that LD was found to decay to the threshold value of $r^2=0.3$ (the generally accepted limit to detect association with a QTL) at 2.20 cM on average, as with DArT markers (Maccaferri et al., 2014). This distance has been used to set a general confidence interval for the GWAS QTLs (=4.40 cM). The LD is therefore estimated to extend at the centiMorgan scale which is in accordance with previous observations for wheat and barley (Muñoz-Amatriain et al., 2014; Maccaferri et al., 2015). The QTL mapping resolution in the panel is thus improved compared with recombinant inbred populations where confidence intervals are mostly within 10-20 cM. The average mapping density of informative SNPs in the Unibo-DP (5.32 SNPs cM^{-1}) is correspondingly high, allowing for nearly complete genome coverage and high confidence in QTL identification. Similarly to maize experiments, QTLs are identified as multiple tightly linked significant SNPs in strong LD and associated with the trait, and rarely as singletons. The outcomes, in terms of power and accuracy of QTL discovery, are considerably better than what was allowed by previous marker technology (SSR and DART; Maccaferri et al., 2014). Accordingly, as compared with the study of Canè et al. (2014), the SNP-based analysis revealed twice as many RSA QTLs as in the previous SSR- and DArT-based analysis.

Phenotypic variation for RSA traits and identification of QTL clusters for early root length and number and their associated effects on GY and TGW

The most well-known and striking example of selection-driven RSA manipulation in wheat is the introgression of the rye/ wheat 1RS.1BL, which is now used worldwide (Howell et al., 2014). In this case, improved adaptation to water stress is due to a deeper root system already detectable at early developmental stages (Sharma et al., 2010; Edhaie et al., 2014; Howell et al., 2014). In other cases, phenotypic variation for RSA observed in RIL populations and diversity panels at the seedling stage (in particular for seminal root length evaluated in rolled germination paper) correlated with measurements carried out on field-grown roots (shovel or extracted soil cores) at vegetative stages up to r=0.6-0.8, while the correlation disappeared at the flowering stage (Watt et al., 2013). However, these observations were not targeted to any specific QTLs. In our study we searched for QTLs expressed at the seedling stage under non-stressed conditions, aiming at identifying constitutively expressed QTLs whose effects are eventually maintained in field-grown plants, at either the vegetative or the reproductive stage. Remarkably, the two mapping populations evaluated in our study showed eight major QTLs for root number and length and three major RGA QTLs with $R^2 > 10\%$.

Linkage and association mapping proved complementary and effective for dissecting the RSA QTLome in durum wheat, highlighting the presence of multiple QTLs, structured into QTL clusters, that could represent either single causal loci with pleiotropic effects on multiple traits or tightly linked loci not resolved by recombination (Tuberosa *et al.*, 2002).

Notably, limited overlap occurred among QTLs found in mapping populations and GWAS, particularly for RGA. This underlines the importance of undertaking both mapping approaches for a more comprehensive investigation of the QTLome. In this context, QTL pyramiding and/or genomic selection approaches involving multiparental populations (Delhaize *et al.*, 2015; Milner *et al.*, 2015) can be envisaged.

QTL clusters for root length and number at the seedling stage are potential candidates for marker-assisted breeding applications aimed at enhancing early rooting capacity. Chromosomes 1A, 1B, 2B, 3A, 3B, 4B, 5A, 6A, 7A, and 7B harboured the most interesting QTL clusters. Prioritizing QTLs for applying marker-assisted selection to enhance crop productivity should be based on the following factors: (i) evidence for significant QTL effects in different genetic materials; (ii) size of the additive effect and R^2 value of the QTL; and (iii) consistency of QTL effects on GY across a broad range of environments (Collins et al., 2008; Tuberosa, 2012). Based on these premises, previous QTL results on GY and TGW in durum wheat have been compared with the results on RSA QTLs reported herein. The QTL clusters on chromosomes 1A, 1B, 3A, and 6B overlapped with QTLs previously reported in durum wheat by Petrarulo et al. (2015) using the Creso×Pedroso mapping population. QTL clusters on chromosomes 1B, 2B, 4B, and 6A were particularly interesting. The RSA OTLcluster 2# (chromosome 1BS, distal region) features QTLs from Co×Ld, Mr×Cd, the Unibo-DP, Creso×Pedroso (Petrarulo et al., 2015), and the bread wheat population Arche×Recital, with all QTL peaks within a 15 cM interval. For this QTL cluster, GWAS showed little or no effect on GY and final TGW. Conversely, based on their concomitant effects on GY and final TGW, the following RSA QTL clusters appear more valuable for breeding purposes: RSA OTLcluster 4# on chromosome 2AL; RSA OTLcluster 5#, 6#, and 7# on chromosome 2B; RSA_QTLcluster_12# on chromosome 4B; and RSA QTLcluster 15# and 16# on chromosome 6A. In particular, RSA QTLcluster 4#, 5#, 7#, 12#, and 15# showed the strongest consistent effects on GY and TGW, mostly in medium- to high-stress environments. The RSA_QTLcluster_5# (chromosome 2BS, distal region) includes QTLs from Co×Ld, GWAS QTLs from the Unibo-DP, and projected QTLs from the hexaploid wheat population Xiaoyan54×Jing 411, which led to the identification of qTaLRO-B1, the only major QTL for RSA in wheat for which a functional basis has been provided (He et al., 2014). Notably, the QTL peaks from tetraploid populations are mostly coincident with the map location of *qTaLRO-B1*.

An additional important region with QTLs for RSA, GY, and TGW was identified on chromosome 2BL (*RSA_QTLcluster_6#*) where eight RSA QTLs were mapped in our durum RIL populations as well as in three hexaploid wheat mapping populations. Notwithstanding the relatively high significance and R^2 values in both RIL populations, this QTL region was not highlighted by GWAS, possibly due to (i)

unbalanced frequencies of the causal alleles in the Unibo-DP; or (ii) small QTL effects that failed to reach significance. This could also have been the case for *RSA_QTLcluster_12#* on chromosome 4BL and for *RSA_QTLcluster_16#* on chromosome 6AL.

QTLs for RGA and their associated effects on GY and TGW

A series of RGA OTLs (~30 in total) were identified. The high heritability of this trait (from 0.65 to 0.90) probably facilitated QTL identification. For RGA, biparental linkage mapping, as compared with GWAS, highlighted a higher number of 'nominal' OTL effects (six in RILs versus four in Unibo-DP), most probably as a consequence of the higher heritability of RGA in RILs as compared with the panel. From a breeding standpoint, the panel provided a valuable estimate of the effects of QTL alleles across many different genetic backgrounds and provides and overview on their allelic distribution in the elite germplasm, two important aspects for their development via marker-assisted selection. Several criteria were considered for prioritizing the RGA QTLs. The striking co-locations between RGA QTLs and both GY and TGW QTL peaks from multienvironmental trials together with the analysis of the relationships ascertained between RGA and GY effects suggest that QRGA. ubo-2A.1, QRGA.ubo-2A.2, QRGA.ubo-2A.3, QRGA.ubo-6A.1, QRGA.ubo-7A.1, QRGA.ubo-7A.4, and QRGA.ubo-7B deserve further characterization, through the development of near-isogenic lines in reference cultivars and/or cumulative/ recurrent selection of the appropriate QTL alleles in segregating populations. Mapping populations allowed us to identify three major QTLs for RGA showing additive effects exceeding 10°, one of which (QRga.ubo-6A.2) was cross-validated in the Co×Ld population and the Unibo-DP. These features make QRGA.ubo-6A.2 a valuable candidate for positional cloning (Salvi and Tuberosa, 2005). The RGA QTLs reported herein for seminal roots can potentially be considered for further development of breeding applications/studies towards the realization (by marker-assisted selection) of the two shallow versus deep root architecture ideotypes schematically depicted in Supplementary Fig. S4.

Out of 30 RGA QTLs, only six were previously described in the wheat germplasm. Thus, 24 novel QTLs for RGA have been identified, including some with large effects such as QRGA.ubo-6A.2. From a breeding standpoint, it should be noted that for most of the RGA QTLs identified in the Unibo-DP, the QTL-tagging alleles for narrow rooting were already present at relatively high frequency in the modern elite germplasm (the SNP allele associated with a narrow angle is present in the germplasm with a frequency >0.50for nine out of 20 QTLs) while the groups of cultivars more directly derived from the local native germplasm showed fixation for the shallow-rooting alleles at several QTLs. Notably, the QRGA.ubo-6A.2 QTL allele for shallow rooting is contributed by Colosseo, a cultivar with drought susceptibility features, particularly under terminal drought/heat stress (as from the Italian durum wheat national network trial reports,

1995–2005) and known to harbour several chromosome regions directly inherited from Mediterranean tetraploid landraces (Maccaferri *et al.*, 2007). Collectively, our findings suggest a gradual change in RSA from native wheat genetic resources (with shallow and densely rooted RSA) to modern wheat cultivars, the latter being characterized by a deeper rooting system more balanced in density throughout the soil depth as shown by Gioia *et al.* (2015) and Nakhforoosh *et al.* (2015).

Conclusions and perspectives

A two-pronged approach based on biparental linkage and association mapping allowed the most comprehensive dissection of the OTLome for RSA in durum and bread wheat at the seedling stage while providing valuable information on the associated effects of these QTLs on grain yield of durum wheat across a range of environments differing widely in water availability and productivity. Comparative analysis with previous QTL studies in durum and bread wheat indicated that a large portion of the QTL clusters for RSA traits described herein (~40% for root vigour and 60% specific for RGA) are novel. On the basis of a multienvironment field study, 15 of the 20 OTL clusters for root vigour overlapped with QTL density peaks for grain weight and/or grain yield. Out of the 30 QTLs for RGA, six did not affect other RSA features, hence providing valuable opportunities for fine-tuning RGA independently of other RSA traits. More importantly, nine RGA QTLs also affected grain yield and, in three cases (QRga.ubo-7B, QRga.ubo-6A.1, and QRga.ubo-7A.4), the QTLs consistently affected both traits across most environments. These three QTLs are promising candidates for positional cloning which would eventually allow for a more effective manipulation of RGA via marker-assisted selection and/or genome editing. The phenotyping of near-isogenic stocks at these three QTLs would also contribute valuable data for modelling the effects of RSA on GY across environments with different water and nutrient availability (Tardieu and Tuberosa, 2010), ultimately allowing for the fine-tuning of the RSA for a more sustainable wheat production.

Supplementary data

Supplementary data are available at JXB online.

Text S1. QTL analysis methodology in the two recombinant inbred line populations.

Text S2. Genome-wide association (GWAS) analysis methodology in the Unibo-DP durum panel.

Text S3. Phenotypic correlations between seminal root and shoot traits.

Table S1. Correlations among root system architecture traits for the Colosseo×Lloyd and Meridiano×Claudio recombinant inbred line populations.

Table S2. QTLs detected for root system architecture (RSA) traits in the Colosseo×Lloyd mapping population.

Table S3. QTLs detected for root system architecture (RSA) traits in the Meridiano×Claudio mapping population.

Table S4. Detailed results of genome-wide association tests for root system architecture (RSA) traits in the Unibo-DP durum panel.

Table S5. Allelic distribution for root growth angle (RGA) QTL-tagging SNPs in the five Unibo-DP durum panel subgroups as defined by population structure analysis. Accessions listed by RGA.

Table S6. Allelic distribution for root growth angle (RGA) QTL-tagging SNPs in the five Unibo-DP durum panel subgroups as defined by population structure analysis. Accessions listed by population structure.

Table S7. Effect on grain yield (GY) and thousand grain weight (TGW) of GWAS QTLs for root growth angle (RGA) in the Unibo-DP panel based on 15 field trials carried out in the Mediterranean Basin.

Fig. S1. Frequency distribution of root system architecture (RSA) traits in the recombinant inbred populations Colosseo×Lloyd and Meridiano×Claudio.

Fig. S2. Root system architecture (RSA) QTLs in tetraploid wheat projected onto an SNP-based tetraploid consensus map.

Fig. S3. Root system architecture (RSA) QTLs, grain yield (GY), and thousand grain weight (TKW) in tetraploid wheat projected onto an SNP-based tetraploid consensus map.

Fig. S4. Schematic ideotypes of wheat with shallow and deep root system architecture as a consequence of different seminal root growth angle (RGA) QTLs contributing towards the ideotypes by marker-assisted selection.

Acknowledgements

We acknowledge the technical assistance of Simona Corneti and Silvia Giuliani in the growth chamber and field experiments, and of Maria Corinna Sanguineti in the statistical analysis. This study was conducted within the framework of the DROPS and EUROOT projects funded by the European Community's Seventh Framework Program under Grant Agreements Nos FP7-244374 and 289300.

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