

To clone or not to clone plant QTLs: present and future challenges

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Recent technical advancements and refinement of analytical methods have enabled the loci (quantitative trait loci, QTLs) responsible for the genetic control of quantitative traits to be dissected molecularly. To date, most plant QTLs have been cloned using a positional cloning approach following identification in experimental crosses. In some cases, an association between sequence variation at a candidate gene and a phenotype has been established by analysing existing genetic accessions. These strategies can be refined using appropriate genetic materials and the latest developments in genomics platforms. We foresee that although QTL analysis and cloning addressing naturally occurring genetic variation should shed light on mechanisms of plant adaptation, a greater emphasis on approaches relying on mutagenesis and candidate gene validation is likely to accelerate the pace of discovering the genes underlying QTLs.

From polygenes to QTL cloning

Classical quantitative geneticists defined 'polygenes' [1] as the many hypothetical genes, with an equally small effect, involved in determining a quantitative trait (i.e. a trait influenced by both multiple genetic and environmental factors). Polygenes have been integrated into most quantitative genetics models; in many cases, these models successfully describe complex phenomena such as the inheritance of quantitative traits, the effect of selection, the consequence of mating behaviour and others. However, within such models polygenes are usually dealt with as a whole, whereas the actual genes remain in what has been defined as a 'statistical fog' [2]. Within this framework, the problem of understanding the molecular nature of quantitative trait variation would have remained unsolved.

The first steps toward resolving this puzzle were based on studies carried out during the first half of the 20th century; these showed that genes with a major effect on quantitative traits do exist and can be experimentally mapped on chromosomes by evaluating the correlation between the quantitative trait value and the allelic states at linked genetic markers [3,4]. This led to the definition of a quantitative trait locus (QTL) as a genetic locus where functionally different alleles segregate and cause significant effects on a quantitative trait. The findings of QTL

studies completed to date indicate an L-shaped distribution of QTL effects (i.e. most QTLs have a small effect and only a few show a strong effect) [5], thus enabling the identification of QTLs with a major effect on the phenotype.

Currently, QTL mapping is a standard procedure in quantitative genetics [6]. QTL mapping usually begins with the collection of genotypic (based on molecular markers) and phenotypic data from a segregating population, followed by statistical analysis to reveal all possible marker loci where allelic variation correlates with the phenotype. Because this procedure only allows for an approximate mapping of the QTL, it is usually referred to as primary (or coarse) QTL mapping. Procedures and strategies for primary QTL mapping are well established [7] and will not be considered here.

More recent technical progress in the area of molecular biology and genomics have made the cloning of QTLs [i.e. the identification of the DNA sequences (coding or non-coding) responsible for QTLs] possible. Here, we present a critical appraisal of the results obtained in this field in plants and discuss the perspectives, with emphasis on several major limitations and promising novel approaches. A literature survey shows that although ~150 research papers reporting original QTL data are published yearly (average of 2000-2004, considering Arabidopsis, soybean, rice, sorghum, maize, barley and wheat), only a handful of studies have reported the cloning of QTLs (Table 1). Besides plants, QTL cloning is rapidly advancing in humans and livestock [8,9] as well as in model species such as yeast, Drosophila, mouse and rat [10–12].

Positional cloning of QTLs

Among the handful of QTLs isolated in plants to date, the majority have been cloned via positional cloning. Positional cloning requires several steps (Figure 1) to enable us to assign a QTL to the shortest possible genetic interval (QTL fine genetic mapping) and to identify the corresponding interval on the DNA sequence (QTL physical mapping) where candidate genes are selected for evaluation. The increase in mapping resolution required by QTL positional cloning is substantial because after primary mapping a QTL is positioned within a chromosome interval of ~10–30 cM, which usually includes several hundred genes (Box 1). Eventually, independent proof is required to validate the role of the identified allelic polymorphism on the observed phenotypic effect.

Table 1. Summary of the main characteristics of the QTLs cloned in plants

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Species	Trait	QTL	Gene	Function	Molecular identification	Candidate gene ^a	R^2 (%) $^{\mathrm{b}}$	Plants $({\sf no.})^c$	Resolution (kb) ^d	ORF (no.) [®]	Identification of QTN	Functional proof	Refs
Arabidopsis	Flowering time	ED1	CRY2	Crypt.	Pos. cloning	Yes (L)	28–56	1822	45	15	Amino acid	Transformation	[71]
		FLW	FLM	#	Pos. cloning	Yes (E)	27	ΝΑ	138	38	Deletion of	Transformation	[31]
	Gluc. structure	GS-elong	MAM	MAM synthase	Pos. cloning	Yes (E)	NA	4600	NA	Ą	whole gene Nucleotide and	No	[82]
	Root morphology	BRX	BRX	ŦF	Pos. cloning	No	80	098	45	10	gene indels Premature stop	Transformation	[83]
Maize	Plant architecture	Tb1	Tb1	TF.	Transp. tagging	Yes (E)	17–31	A N	AN	A A	No, possibly	Complementation	[33,42]
Rice	Heading time	Hd1	Set	⊭:	Pos. cloning	Yes (L)	67	1505	12	2	regulatory No	Transformation	[84]
	Heading time Heading time	Hd3a Hd6	Hd3a ¤CK2	Unknown Protein kinase	Pos. cloning Pos. cloning	Yes (L) No	A A	2207 2807	20 20	4 –	No Premature stop	Transformation Transformation	[30]
	Heading time	Ehd1	Ehd1	B-type response	Pos. cloning	No	Ą	> 2500	16	က	codon Amino acid	Transformation	[82]
Tomato	Fruit sugar content	Brix9-2-5	Lin5	Invertase	Pos. cloning	Yes (L)	NA	7000	0.5	_	Amino acid	Complementation	[86,87]
	Fruit shape	Ovate	Ovate	Unknown	Pos. cloning	No	48-67	3000	55	8	Premature stop	Transformation	[88]
	Fruit weight	fw2.2	ORFX	Unknown	Pos. cloning	No	30	3472	92	4	Unknown regulatory	Transformation	[58,89]
											Variant		

Abbreviation: crypt., cryptochrome; gluco structure, glucosinolate structure; MAM synthase, methylthioalkylmalate synthase; NA, not applicable or not available; ORF, open reading frame; pos. cloning, positional cloning; OTL after primary QTL analysis; (L) indicates late evidence, after physical mapping and/or sequencing quantitative trait locus; QTN, quantitative trait nucleotide; TF, transcription factor; transp. tagging; transposon tagging. Evidence

^bProportion of phenotypic variance explained by the QTL in ^cDimension of the population used for fine mapping.

DIMENSION Of the population used for the mapping.

DNA physical interval completely linked with the QTL.

Number of ORFs completely linked with the QTL.

Types of genetic material suitable for QTL fine mapping

A widely adopted strategy to estimate the position and effect of a coarsely mapped QTL more accurately is to create a new experimental population by crossing nearly isogenic lines (NILs) that differ only in the allelic constitution at the short chromosome segment (usually varying from ~10 to 30 cM in length) harbouring the QTL (QTL-NILs). In such a population, because of the absence of other segregating QTLs, the target QTL becomes the major genetic source of variation, and the phenotypic means of the QTL genotypic classes (+/+, -/-) and, when present, -/+) can be statistically differentiated and genotypes recognized accordingly. Appropriate replication and/or progeny testing are generally implemented based upon the heritability of the trait considered. Under such conditions, the QTL is considered Mendelized [13], and cM distances between a QTL and the nearby molecular markers can be estimated more precisely.

Beginning with the same population in which primary mapping was carried out, QTL-NILs can be produced by (i) marker-assisted backcross introgression (i.e. substitution) of one QTL allele into one or both parental genetic backgrounds or (ii) iteratively identifying and selfing individuals that are heterozygous at the QTL region. QTL-NILs can also be efficiently identified within introgression libraries (ILs) (i.e. collections of lines where each line is isogenic to a background parental line, with the exception of a single short chromosome segment introgressed from a donor) [14]. Within an IL, the donor genome should be completely represented among the different IL lines, therefore, potentially a QTL-NIL exists for any segregating QTL. Remarkably, the same tomato IL [15] provided the source of QTL-NILs used for cloning three tomato QTLs. NILs suitable for positional cloning can also be produced using the advanced backcross QTL analysis (ABQA) method, which combines backcrossing chromosome segments from a wild accession within an elite line coupled with some level of phenotypic selection against extreme phenotypes [16]. Crucial aspects to be considered are the time and effort required for developing collections of IL or ABQA lines, as well as the limited genetic variability as a result of using only two parental lines: no matter how carefully the parental lines are chosen, the collections will only segregate for a fraction of the many more QTLs segregating for the same trait in other populations. This important limitation can be partially overcome through the use of multiparental intercrossed populations [17,18]. These populations are generated by crossing a carefully chosen set of parental lines capturing much of the genetic variation of the species, followed by several cycles of intermating. This approach should increase the efficiency of QTL mapping both in terms of detection (segregation is expected at many loci) and genetic resolution (many rounds of meiosis). A substantial increase in genetic resolution can also be obtained by intercrossing standard biparental populations [19].

During the fine-mapping step, the resolution of the target QTL in two or more linked loci can bring positional cloning projects to an end when the proportion of phenotypic variability explained by each QTL is too small to be revealed with a realistically manageable number of

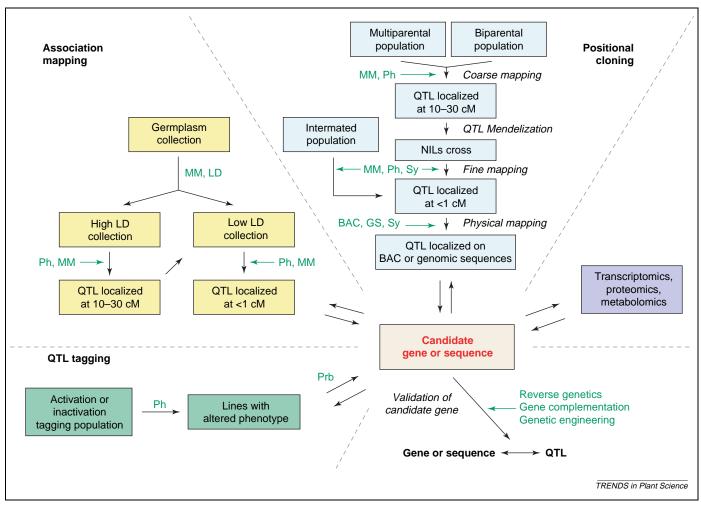


Figure 1. Flow-chart depicting the molecular dissection of quantitative traits by positional cloning, association mapping and QTL-tagging. Boxes indicate starting materials or major milestones. Major experimental processes are indicated in black italic font. For each experimental process, relevant genetic, molecular or analytical tools are indicated in green font. Abbreviations: BAC, library of bacterial artificial chromosomes; GS, genomic sequence; LD, data on linkage disequilibrium; MM, molecular markers; NlLs, nearly isogenic lines; Ph, phenotyping; Prb, probing with tagging agent, for example, transposon; Str, data on population structure; Sy, synteny. Reverse genetics includes transposon and T-tagging, activation tagging, TILLING and RNAi.

replications. QTL clusters have been observed in plants [20–22]; however, cloning was accomplished when one of the linked QTLs retained most of the effect [23,24].

Molecular genotyping

The recruitment of polymorphic markers required for fine mapping a QTL is fairly simple for *Arabidopsis* and rice because the whole genome has been sequenced, and also for species such as maize or tomato for which genomic sequencing is under way or information is available in terms of, for example, ESTs and BAC ends (See Glossary) [25,26]. However, in species for which detailed sequence

Box 1. How many genes underlie a primary QTL?

It is impossible to identify the gene subtending a specific QTL after its mapping by primary analysis because of the poor resolution of the analysis itself. Primary analysis usually maps a QTL within a chromosome region (known as QTL supporting interval) of $\sim 10\text{--}30$ cM. Using current estimates of the total gene number and the genetic and physical length of Arabidopsis and maize genomes (available at http://www.arabidopsis.org/ and http://www.maizegdb.org/), it can be calculated that a 10-cM chromosome interval on average corresponds to ~ 2.1 Mb and 440 genes in Arabidopsis or to ~ 12.4 Mb and 310 genes in maize.

information is not available or cannot be deduced from syntenic relatives, many molecular markers (e.g. amplified fragment length polymorphisms) need to be screened in genotypes contrasted at the target region (e.g. pair of QTL-NILs). Synteny with *Arabidopsis*, *Medicago* or *Lotus*, and rice should assist in identifying additional markers for cloning projects within *Brassicaceae*, legumes and cereals, respectively [27]. With regards to genotyping techniques, microarray-based platforms appear to be particularly promising for high-throughput identification of polymorphisms [single nucleotide polymorphisms (SNPs) and indels] at thousands of loci [28,29].

Physical mapping and the identification of candidate sequences

When the genetic resolution approaches the cM level, the markers closest to the QTL are used for anchoring the genetic map to the physical map (i.e. the genomic sequence or a BAC contig covering the QTL region). Early transfer of the information to the physical map enables the efficient generation of new single-copy markers useful for refining the genetic mapping and for searching candidate genes. Even if only a BAC contig is available, sequenced BAC

Glossary

BAC contig: a contiguous set of overlapping BAC clones.

BAC ends: portions of a BAC genomic insert that are near the cloning site of the vector. BAC ends can be easily sequenced and used to anchor the BAC clone to other sequences or to a genetic map.

BAC (bacterial artificial chromosome): a plasmid vector capable of 100-150 kb

Candidate gene: a gene that based on its map position and/or its nucleotide sequence could be responsible for a given phenotype.

cM (centiMorgan): indicates the distance of two loci on a chromosome based on the observed or estimated frequency of crossovers over the total number of scored meiosis. For low values, one cM corresponds to 1% of gametes carrying a crossover chromosome.

EST (expressed sequenced tag): short (usually <500 bp) sequence, usually obtained by raw sequencing of a random cDNA. Collections of ESTs provide a quick way to represent a substantial portion of the gene complement of a

Haplotype: a combination of alleles at different loci on the same chromosome seament.

Kb/cM rate: the average chromosome distance (in base pairs of DNA) per unit of genetic distance (cM). It can refer to a general mean of a species (computed as genome dimension in kb/total genetic map) or to a specific chromosome region.

LD (linkage disequilibrium): the level of non-random assortment of alleles at different loci.

Molecular marker: a locus whose genotype can be inferred by a molecular assay based on more or less direct analysis of the DNA sequence

Phenotypic variance: a statistical index of the variability of the population. It includes two major components; variability produced by genetic segregation and interaction of QTLs and variability produced by random environmental

QTN (quantitative trait nucleotide): the DNA sequence polymorphism responsible for the QTL effect.

RNAi (RNA interference): a biological process (and a reverse genetics technique) causing post-transcriptional inhibition of the expression of a target gene by the action of small (21-26 nucleotides) RNA molecules.

Segregating population: a population of individuals differing in allelic constitution at one or more genetic loci.

SNP (single nucleotide polymorphism): a point mutation that can be targeted by molecular techniques and can thus be exploited as a molecular marker.

Somaclonal variation: the de novo variation observed in plants regenerated from tissue culture. It can be caused by transposon activity, late replication of heterochromatin, gene amplification, mitotic crossing-over and/or changes in methylation that occur during the in vitro phase.

Synteny: conservation of genome organization (including the linear order of genes on chromosomes, which is called collinearity) among phylogenetically

TILLING (targeting induced local lesions in genomes): a reverse genetics technique that enables individuals carrying point mutations at a target gene of known sequence to be identified within a chemically mutagenized population.

ends can often be transformed in genetic markers, and low-pass shot-gun sequencing can provide a glimpse of local gene content. In this phase, bioinformatics provides an important contribution in terms of gene prediction and annotation, and exploitation of syntenic relationships.

Among the studies considered here, only two managed to reduce the number of genes co-segregating with the target QTL (Table 1) to one. In one case, Hd6, the QTN (i.e. the quantitative trait nucleotide polymorphism responsible for the QTL effect) was a nucleotide substitution in one of the two alleles that caused a premature stop codon [30]. When the physical region co-segregating with the QTL includes more than one gene (can be up to 38; [31]), candidates can either be identified via function prediction and selected for further testing (see Cry2, FLM, Hd1 and Hd3a in Table 1) or they might not be clearly evident. When multiple coding sequences with no obvious candidate gene are identified, two possible options are to increase the mapping resolution or to test each open reading frame (ORF) functionally.

Validation of a candidate gene or sequence

The functional testing of a candidate gene(s) can be performed by overexpressing or down-regulating the target gene through genetic engineering or RNAi [32], or by genetic complementation of a known mutant [33]. If available within the species under investigation, reverse genetics tools such as T-DNA or transposon-tagged populations [34] and/or TILLING [35] can also be exploited. Compared with transposon tagging, TILLING and RNAi are appealing alternatives for their almost universal applicability and for providing subtle changes of gene functionality comparable to those observed naturally. Gene replacement, still in its infancy but already reported in rice [36], is the ultimate tool for validating candidate genes.

The validation of QTNs in non-coding regions is one of the current major challenges. Regulatory regions close to (e.g. promoters) or far (e.g. enhancers or silencers) from the regulated gene are likely to host sequence polymorphisms causing variation in quantitative phenotypes. It can also be predicted that QTLs will be found at microRNA loci and at regions controlling chromatin methylation and/or organization (e.g. folding). Furthermore, transposon insertions have already been shown to be responsible for changes in gene expression [37]. However, for most of these genomic features a structural characterization is still lacking, thus hindering the recognition of their role in the control of quantitative traits. Regulatory elements at several tens of kb from coding sequences have already been shown to act as QTLs [38-40], even though the causal QTNs have not yet been identified.

After considering all the above-mentioned aspects, it is clear that positional cloning of QTLs in plants remains a demanding and daunting undertaking. Furthermore, positional cloning has been limited exclusively to major QTLs because all the cloned QTLs showed an R^2 value higher than 15% in the primary genetic analysis (Table 1). R^2 values can be grossly under- or overestimated [41] because of statistical artefacts and because epistasis can modify the genetic effect of the target QTL when the genetic background changes [42], for instance during QTL-NIL production. Therefore, an independent evaluation of the QTL effect (e.g. by developing and phenotypically testing QTL-NILs [43]) is recommended before embarking on QTL positional cloning.

Cloning QTLs by association mapping

As an alternative to positional cloning, QTLs can be molecularly resolved through association mapping [44] (i.e. by identifying, within a set of genotypes such as germplasm accessions and cultivated varieties, a statistical association between allelic variants at marker or candidate loci and the mean of the analysed trait). The analysis evaluates the trait mean change caused by the substitution of one allele with another. For QTL cloning in plants, the interest lies in (i) the possibility of finding chromosome regions important for controlling quantitative traits without the costly and time-consuming production of large experimental populations [45], (ii) the potentially high genetic resolution provided by the many meiotic events that occurred during past generations, and (iii) the possibility of surveying many functionally diverse alleles per locus.

A major factor to be considered in association mapping is the level of linkage disequilibrium (LD) among the tested accessions. In plants, the extensive LD analyses conducted in Arabidopsis and maize [46] have indicated that whereas LD persists over hundreds of kb in *Arabidopsis*, in maize, LD decays after a few kb, although it can extend significantly farther in collections of elite germplasm [46]. With high LD values, marker-trait association can theoretically be revealed with a manageable number of molecular markers. In this case, the expected mapping resolution will only be sufficient for the discovery and coarse mapping of the QTL. However, when testing germplasm panels with low LD, the diagnostic power of a single marker will only extend a short way and thus a prohibitively high number of markers would be required for a whole genome scan. In this situation, association mapping can still be used to fine map the QTL at the gene level after the QTL is positioned using standard mapping procedures. Based on this, it is conceivable that different sets of genotypes, characterized by high or low LD, can be assembled and used for QTL discovery or candidate gene validation, respectively, as has been suggested for human genetics [47]. Population structure (i.e. the possible presence of hidden subgroups, because of e.g. relatedness and selection, with an unequal distribution of alleles) might influence the efficacy of this approach by causing spurious trait-marker associations [48].

EcoTILLING [49] is a powerful approach for identifying different haplotypes (combinations of allelic variants) at target loci and making them available for association mapping, enabling the identification of virtually all SNPs and small insertion or deletions within a ~1-kb window in a set of genotypes at a fraction of the sequencing cost. However, the necessity of also screening regulatory regions that are often distant from the effector genes indicates that selecting the candidate sequences to be tested for association mapping is not a trivial task if the genomic scan aims to be comprehensive. Examples of identifying associations between haplotype variation at a candidate gene and a quantitative trait have been reported in *Arabidopsis* [50], *Brassica* [51,52], potato [53] and maize [54–56].

The identification of a statistically significant association between haplotype variation at a candidate gene or sequence and a quantitative phenotype should be followed by validation experiments similar to those used in the positional cloning approach previously described.

Functional genomics and QTL cloning

The use of functional genomics is contributing to many aspects of QTL analysis and cloning. Transcriptional profiling between contrasting QTL genotypes can quickly provide a list of genes differentially expressed; subsequently, those genes functionally related to the target trait and mapping at the QTL region can be selected as candidates [57]. Unfortunately, the number of QTLs cloned to date in plants is too small to test the validity of this approach. Indeed, when the QTL has been shown to involve a difference in gene expression level between

alleles, the difference was either too low (approximately twofold [33]) or showed too strong a spatial and/or temporal pattern [58] to enable them to be identified using standard microarray-based transcriptome analysis. Other profiling platforms, such as SAGE (serial analysis of gene expression [59]) and MPSS (massively parallel signature sequencing [60]) are better suited to detecting subtle differences in gene expression. Transcript profiling can reach the sub-tissue level of resolution if carried out in combination with laser-capture microscopy [61].

The expression profiling of a mapping population at the mRNA or protein level enables us to treat the level of expression of a single gene as a quantitative trait and to dissect its genetic control by QTL analysis [5,62–64]. The loci controlling the level of gene expression have variously been named transcript quantity loci (TQLs), expression QTLs (eQTLs) or protein quantity loci (PQLs) [64,65]. Correspondence between eQTLs and/or PQLs and candidate genes with QTLs for morpho-physiological traits has already been observed in small- or medium-scale experiments [66-68]. Microarray-based studies have mapped eQTLs at the same location of the gene whose expression was measured, thus indicating a role for *cis*-regulatory allelic variation, and also at distant chromosome positions [65,69]. The same studies highlighted the presence of eQTL 'hot spots' (i.e. chromosome regions apparently responsible for controlling the simultaneous expression of many genes).

Adding a new framework: mutations at QTLs and QTL tagging

Donald Robertson [70] suggested that qualitative mutant alleles and wild-type alleles at loci affecting quantitative traits are the extremes of a possible range of effects, with QTLs resulting from the segregation of naturally available alleles with milder effects. Robertson's hypothesis was confirmed in those cases where a mutant was available for the gene subtending the target QTL [42,71]. Along this line, it was recently argued that mutagenesis could be more efficient for dissecting the genetic basis of quantitative traits than is QTL analysis, which only provides 'accidents of history' allelic variants [72]. A direct method for identifying such genes would be to use a tagging (insertional) approach. Such a framework would require the phenotypic screening of an insertionally mutagenized population for the target quantitative trait to identify those lines with a phenotypic mean value outside a predicted range because of environmental effects [70,73]. The complete screening experiment would involve a manageable number of plants (e.g. up to 20 000–30 000) if multiple insertion systems are used and several quantitative traits are concurrently evaluated [73]. The gene functionally modified or inactivated by the insertional agent could be rescued using standard molecular procedures. Following a similar approach, QTL tagging has already been successfully accomplished in Magnaporthe [74], the causal agent of rice blast, and in Drosophila [75]. In plants, QTL tagging could be carried out using several different approaches, based on T-DNA as well as DNA-transposons and retrotransposons. However, systems relying on callus cultures (e.g. activation of rice TOS-17 retrotransposon [76]) should be considered with caution because of the occurrence of somaclonal variation [77], which can potentially alter any quantitative trait and therefore hinder the identification of the tagged QTL. Instead, other interesting resources are the Ac–Ds-based insertional populations developed in rice [78]; following the introduction of heterologous transposons, the majority of mutational events are created by new transposition activity. In maize, a Mu-based insertional population has been developed in a non-segregating genetic background [79]; within such a population, most of the quantitative variability can be attributed to the segregation of the tagged QTLs.

Increasingly important role of candidate genes

Classically, a link between a gene and a quantitative trait can be hypothesized based on linkage information (all genes co-segregating with a QTL are positional candidate genes) or commonality between the quantitative trait physiology and the biochemical function of the gene (functional candidate gene) [80], or both. For instance, completion of genome sequences and improved bioinformatics should facilitate in silico cross-matching of candidate sequences with QTLs in programmes of positional cloning or association mapping. The creation of more powerful bioinformatic tools for gene annotation should facilitate the choice of functional candidates among and outside the positional candidate genes [81]. In addition, a better understanding of the mechanisms behind the regulation of gene expression should extend the concept of candidate gene to include cis-acting regulatory sequences. Therefore, in the future, it is conceivable that QTL cloning will increasingly rely on candidate gene information and that this will be made possible by exploiting the available reverse genetics tools (Figure 1).

Conclusions

The QTLs cloned to date in plants are likely to represent a biased sample of those governing the variability of target traits; only major QTLs, mostly identified in wide crosses (e.g. indica × japonica rice subspecies, teosinte × cultivated maize and wild×cultivated tomato), have been successfully targeted. However, the constant improvement of the molecular platforms, new types of genetic materials, progress in bioinformatics and the increasing availability of tools and platforms for functionally testing candidate genes should facilitate QTL cloning and offer the opportunity of targeting QTLs other than those with a major effect. Quantitative approaches to dissect mutants genetically will also have to be extended if more subtle regulators of complex phenotypes are to be uncovered.

From a breeding standpoint, we need to continue investigating natural variation to unlock the allelic richness present in germplasm collections, which should enable us to use marker-assisted selection and genetic engineering more effectively to introduce valuable alleles in crops. At the same time, QTL cloning should improve our understanding on how nature shaped genetic variability during adaptive evolution.

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