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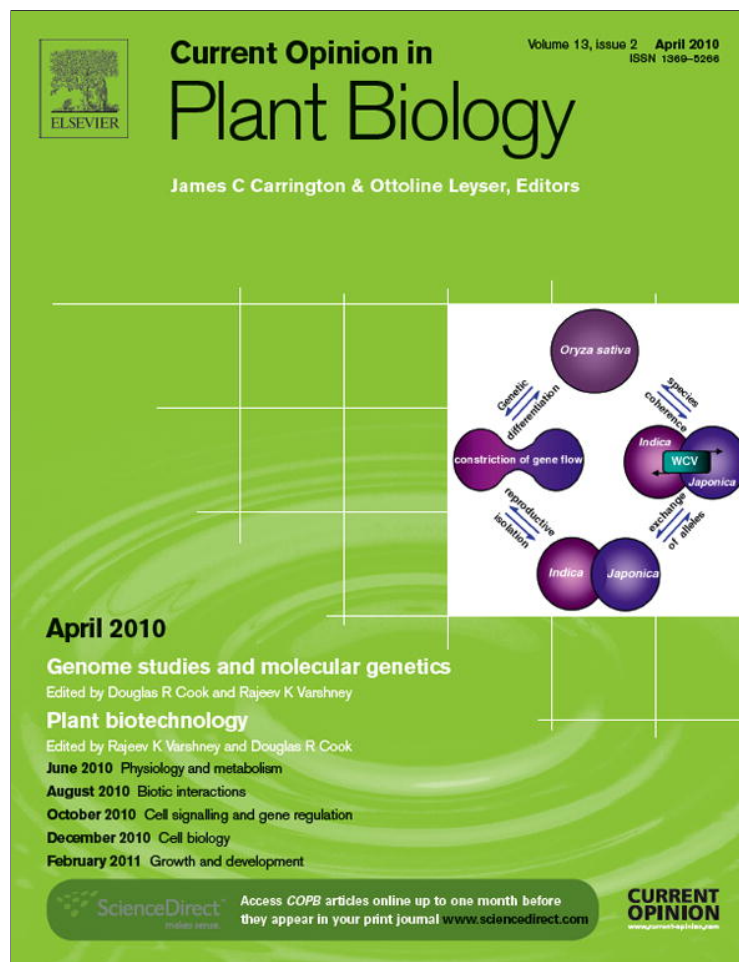
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Dissection and modelling of abiotic stress tolerance in plants

 François Tardieu¹ and Roberto Tuberosa²

Plants have acquired a variety of whole-plant protection mechanisms in response to abiotic stresses, often at the expenses of performance. Hence, a given trait can have positive, negative or no effect depending on the stress scenario. A new approach has emerged that dissects yield and integrative traits that influence stress tolerance into heritable traits (e.g. sensitivity parameters or architectural traits) by using phenotyping platforms with model-assisted methods. The genetic and physiological mechanisms accounting for the variability of these traits and their effects on yield are considered in a second step. Effects of traits on yield are analysed via a combination of modelling and field experiments, which allows identification of the stress scenarios where a given allele has favourable effects.

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Introduction

Tailoring genotypes with acceptable performance under drought, high temperatures or low nutrient availability is essential for the sustainability of crop production in view of climate change [1–3]. Conventional breeding has been successful [4,5^{••}], but the pace of genetic progress must now increase to meet the projected demand for agricultural products [3]. This is a challenging endeavour because it involves a plethora of loci (mostly quantitative trait loci: QTLs) whose effects differ in the many possible stress scenarios [6^{••}]. In field trials encompassing a broad range of pedoclimatic scenarios, a given QTL has at best shown a significant effect on yield in only half of the tested environments [6^{••},7^{••}].

The most common strategy has been to identify genes involved in mechanisms of tolerance with a large effect that overrides this complexity. This has led to the isolation and ectopic expression of master genes able to fine-tune the

plant reactions to stress via transcription factors, hormonal balance or metabolic protections [8,9]. Although success stories have been reported on the use of such genes to tailor drought-tolerant genotypes [9,10], this approach has not yet led to the release of new cultivars.

Alternative strategies involving the use of natural genetic diversity under diverse environmental conditions require much larger numbers of studied lines and of environmental scenarios. Until recently, it was nearly impossible to perform physiological analyses of such complex situations, but progress in different disciplines opens new avenues: (i) High-throughput phenotyping now allows reproducible measurement of drought-related physiological traits in hundreds of genotypes, in controlled conditions and in the field [11–13]. (ii) Methods are available to describe mathematically traits that vary rapidly with environmental conditions (e.g. growth or gas exchanges), via model-assisted phenotyping [14^{••},15] and meta-analyses of large datasets [16[•],17^{••}]. (iii) The multiplicity of stressing environments, difficult to handle experimentally, can now be addressed owing to the progress of models [6^{••},18,19^{••}]; if a model adequately describes the effects of the genetic variability in a few climatic scenarios, it can then be extended to a much larger number of scenarios in order to evaluate the comparative advantage of a given allele in different environments [19^{••}]. (iv) Methods for statistical analysis of the genotype × environment × management interaction have also enormously progressed, thereby allowing relevant analyses of large networks of field experiments [5^{••},20[•],21].

We review here the implications of this recent progress on the strategy for improving drought tolerance. A two-step strategy is presented in which, first, phenotyping platforms facilitate the genetic dissection of yield into simpler and more heritable traits, associated with QTLs that control their expression. The effects on yield of this genetic variability are then scaled up in a second step, via a combination of modelling and field studies.

Why model? Selection for stress tolerance an optimisation of contradictory processes

Tolerance to abiotic stresses is an ambiguous concept, even after distinguishing different strategies (e.g. escape, avoidance and tolerance) [22]. In this review, we shall not consider traits (e.g. flowering time) that provide an escape mechanism. According to the nature, timing and intensity of the stress episode, tolerance can be defined either by the ability of plants to survive severe stresses and complete their cycle, or to achieve acceptable yield under a milder stress. When the metabolic cost of protection

mechanisms is low, these two definitions can be congruent. For example, the mechanisms that mitigate the negative effects of heat stress (e.g. via the protection of photosystems and of other metabolic processes) may not be very different from those permitting high biomass accumulation at high temperatures [23,24]. Conversely, protection mechanisms against water deficit are most often physiologically associated with a reduction of biomass accumulation, hence yield potential [15,25[•]]. Biomass accumulation and transpiration are intrinsically linked via their main determinants, that is, stomatal aperture and leaf area [26], so there is a 'built-in' contradiction between biomass accumulation and stress avoidance via transpiration. Breeding can only optimise the terms of this contradiction, for example, by increasing the ratio of accumulated biomass to transpiration (water-use efficiency, WUE) [27–29] or increasing the degree of root or leaf growth maintenance under water deficit [30–32].

Modelling may well provide the most efficient way to find optimums if the effect on yield of a given trait is a trade-off between positive and negative effects, with different optimums depending on the prevailing environmental conditions [6^{••},19^{••}]. A striking example of such complexity is the genetic improvement of WUE. The introgression into elite material of genomic regions that confer high WUE has led to a clear yield improvement of wheat when subjected to a severe soil water deficit [33^{••}]. Conversely, selection for increased WUE has frequently led to small plants with reduced transpiration, biomass and yield in less severe drought conditions [25[•],27]. The genetic variability of root systems is another example. Improving root systems clearly increases yield under water deficit in species with weak root characteristics such as rice [32] but can have more complex and unexpected consequences otherwise. QTL alleles enhancing root architectural features influenced positively maize yield in a particular genetic background [34[•]] but failed to do so in a different one [35]. Furthermore, breeding programmes aimed at improving maize yield under drought conditions have actually led to a decrease in the weight of the root system [36,37]. A high allocation of biomass to roots can influence positively yield in a soil with water stored in deep layers but not under limited soil water reserve, owing to a poor return in terms of water uptake of the carbon invested in root growth and metabolism. A similar conclusion has been drawn for the relation between root density and nitrogen uptake [38].

Dissection of complex phenotypes into more simple and heritable traits

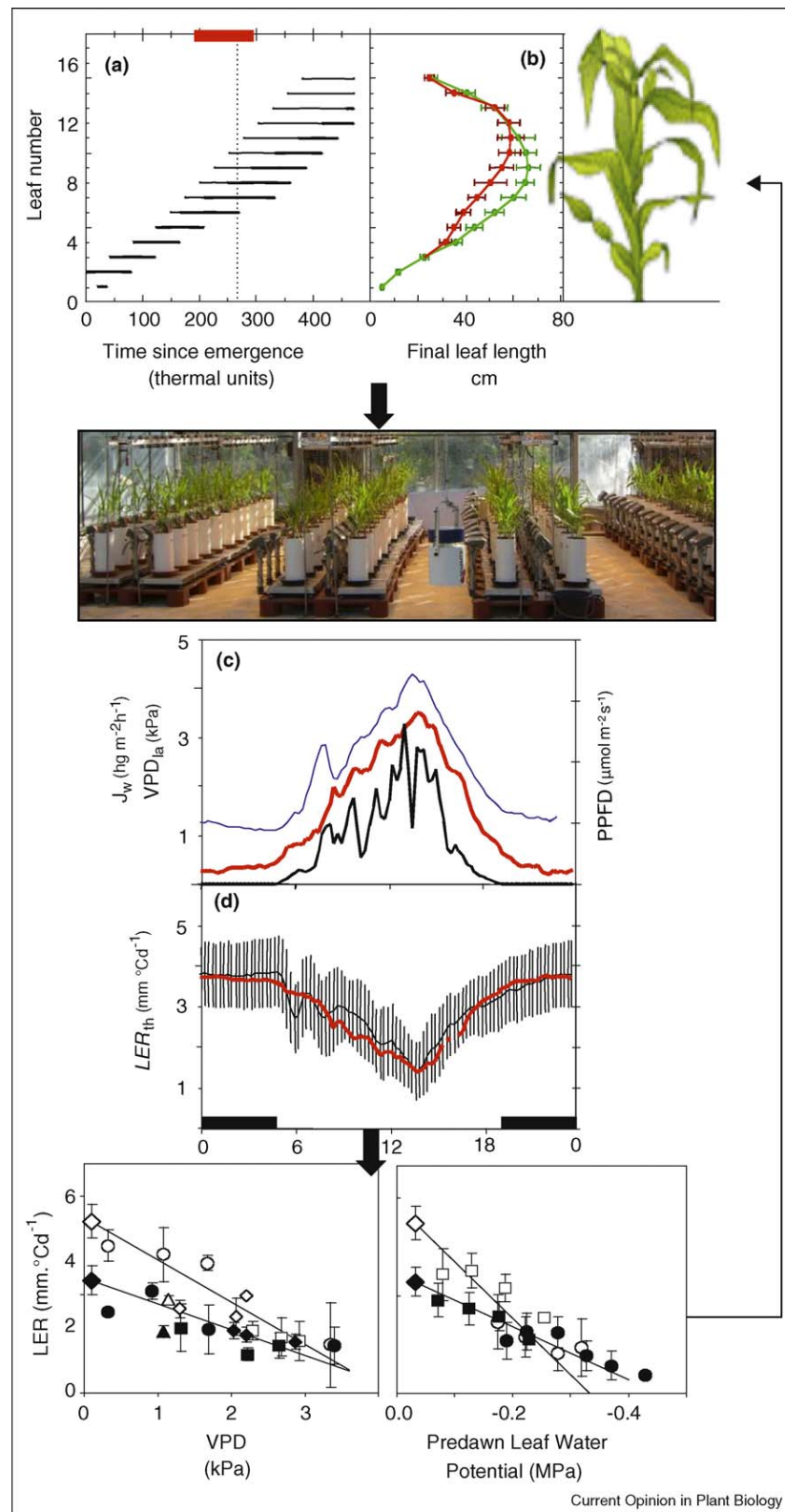
The consequence of the studies reviewed above is that any trait involved in stress tolerance can be considered as having positive, negative or no effect on yield depending on the stress scenario. It follows that the genetic variability of those traits should be analysed and dissected *per*

se, independently of its consequences on yield in a given environment [6^{••},15].

An example of dissection is presented in Figure 1, which synthesises recent studies in maize [13,30,31^{••},39,40] and rice [16[•],41]. The development of leaf area was dissected into that of leaves that grow at different periods of the plant cycle and, consequently, under different environmental conditions. The beginning and the duration of each phase can be modelled as a function of temperature-compensated time for a given genotype [40–43]. Convergent experimental evidence shows that a temporary water deficit essentially affects the growth of leaves which develop during the stress [40] (Figure 1). It is therefore possible to dissect plant leaf area into single leaves, whose individual size will depend on the environmental conditions during their development. Because individual leaf size has usually a low heritability under stress [44], its genetic analysis required further dissection. It was dissected into a duration of development (with a low effect of environment except temperature) and leaf growth rate (highly sensitive to environmental conditions) which can be measured in an automated phenotyping platform [13,41] (Figure 1). While leaf elongation rate varies with time according to environmental conditions, its maximum value observed during the night and its response to evaporative demand or to soil water deficit are common to different experiments for each individual genotype of a mapping population, thereby permitting identification of stable QTLs of sensitivity to environmental variables. A similar approach has been proposed for the dissection of the sensitivity of flowering time to environmental conditions [14^{••},45].

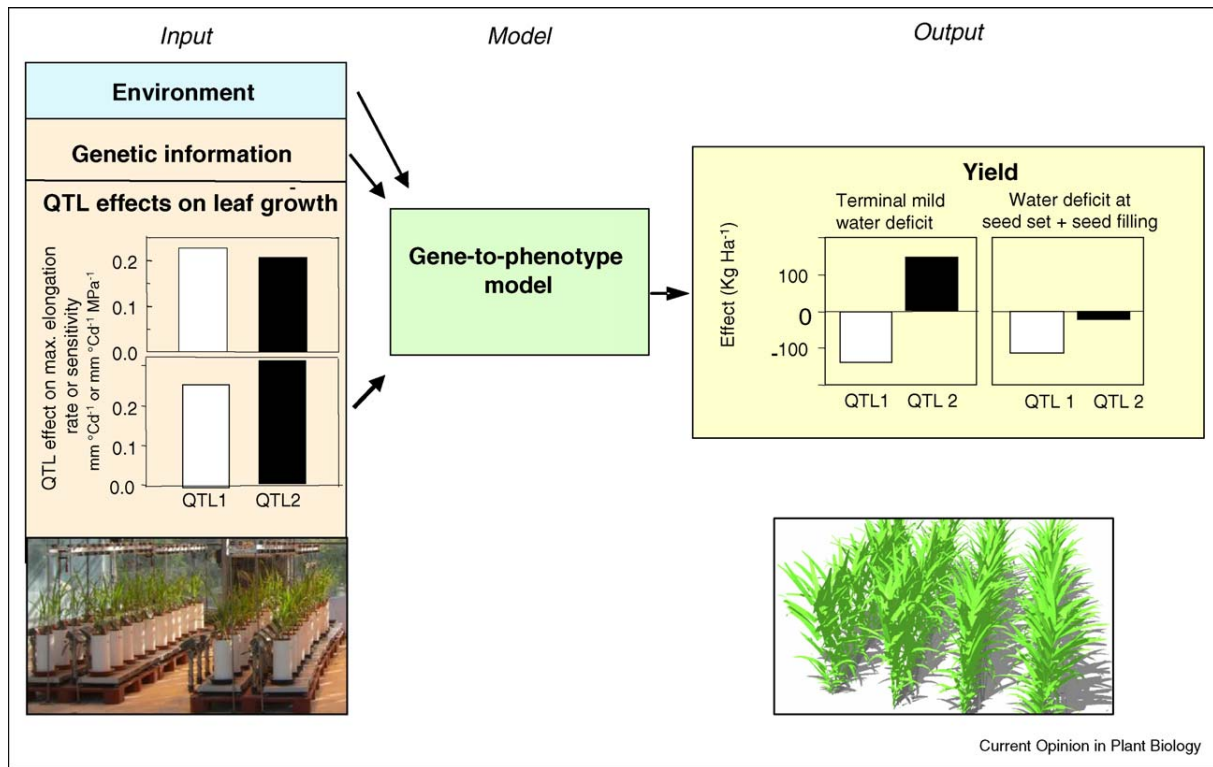
The dissection approach reviewed above differs from that used in several breeding programmes on the basis of selection for secondary traits correlated to yield [29,46]. Nonetheless, common features of both approaches are the search for higher heritability and for a genetic link to yield. A first major difference is that the traits in a dissection approach are most often 'hidden parameters', not directly measurable, which reproducibly characterise studied lines. They can be, for example, sensitivity parameters detected in a regression analysis [14^{••},30], architectural features [47] or temperature-compensated leaf appearance rate [40]. A second difference is that the dissection analysis is intimately associated with the model-assisted scale up (see below). Hence, target traits are chosen in such a way that they can be explicitly related to biomass accumulation and yield via equations accounting for environmental conditions [26], which relate biomass accumulation to the amount of light received, the proportion of incident light intercepted by leaves (dependent on leaf area), the efficiency of transformation of intercepted light into biomass (dependent on photosynthesis rate) and the duration of the plant cycle. It is noteworthy that this was the case in the

Figure 1



An example of dissection: plant leaf area in maize. **(a)** (Field) Successive leaves develop following a strict programme in terms of thermal time [39]. **(b)** If a stress occurs during a limited period of time (horizontal red bar, in (a)), only leaves which develop during stress are affected in size (green, well watered; red, stressed). The plant leaf area can therefore be modelled as a succession of independent leaves [40]. **(c)** (Phenotyping platform). Leaf

Figure 2



Simulation of the effects on yield of two QTLs involved in leaf elongation rate and its sensitivity to water deficit (modified from [19**]). The simulation is based on climatic data over 45 years of non-irrigated fields and two types of soil. The two QTLs have similar effects on the maximum leaf growth rate (upper panel, *inputs*), but different effects on sensitivities of leaf growth (lower panel) and Anthesis–Silking interval to water deficit. Genotypic data of each recombinant inbred line, QTL effects and environmental data of each year are fed into the model, which then simulates yield in each situation. The effects on yield of alleles at the two studied QTLs are then calculated, and are presented in the right panel in two clusters of pedoclimatic scenarios, either with a favourable soil water status during most of the crop cycle or with a water deficit from flowering time onwards (two additional clusters of situations are not presented here). Note that positive alleles for both elongation rate and sensitivity translate to either positive or negative alleles for yield depending on QTL and environmental conditions. Inputs: allelic effects on maximum leaf growth (upper panel) or on sensitivity of leaf growth to soil water potential (lower panel) a positive effect indicates a greater sensitivity. Virtual field representation in ‘outputs’, courtesy C. Fournier.

example presented in Figure 1, in which traits measured in a few fields and in a phenotyping platform allowed prediction of leaf area and biomass accumulation in field experiments in Mediterranean and temperate environments in France and Australia [40].

Which levels or type of dissections are relevant to abiotic stress tolerance? The answer largely depends on the heritability observed at each step of the dissection. Heritability is an essential feature for the success of genetic analyses. The high heritability of flowering time [48] has allowed cloning of major QTLs in rice [49] and maize [50], while no QTL of root architecture has been cloned

to date, mainly owing to the usually low heritability of root traits [51]. Unexpectedly, when a rice mapping population was evaluated under managed-stress conditions, yield itself showed the highest heritability compared with secondary traits even in dry conditions [52,53]. This can be partly attributed to the fact that a single, well-controlled water deficit was applied at flowering time in all experiments, thereby resulting in highly reproducible results that apply to one climatic scenario only. Further dissection probably has no major interest in this case. Conversely, the approaches reviewed above, which addressed more diverse climatic scenarios, allowed the genetic dissection of an integrative phenotype with low

(Figure 1 Legend Continued) elongation rate of an entire mapping population is measured with a 15-min definition, together with transpiration (J_w , red line) and environmental conditions (light intensity, PPFD, black and leaf-to-air vapour pressure deficit, VPD_{la} , blue). (d) Leaf elongation rate (black line) is at each time related to the reciprocal of transpiration rate (red line). Vertical lines, interval of confidence [62]. (e, f). Multiple experiments in greenhouse and growth chamber allow compilation of response curves which reflect the sensitivity of leaf elongation rate to (e) evaporative demand of well-watered plants during the day and (f) to soil water potential, of drought plants during the night [31**]. Upward arrow: Leaf area in the field can be predicted from the time course of environmental conditions, the model of development in (a) and the sensitivity parameters in (e) and (f) [40].

heritability into a small number of features with higher heritability [14^{••},30,31^{••},44].

Finally, it is noteworthy that this dissection step does not require full understanding of the physiological processes that explain the considered traits. In the case of sensitivity of leaf growth to water deficit, the genetic variability might be associated with mechanisms as different as aquaporin activity, abscisic acid effects, cell cycle or cell wall properties, which often show low heritability owing to compensation mechanisms that differ between lines [54–56]. If the selected traits are stable characteristics of a genotype, they can be considered as ‘meta-mechanisms’ at the whole-plant level [15], managed experimentally for QTL detection and used in modelling.

Scaling up: from QTLs of simple traits to crop performance under stressed conditions

The identification of loci (genes or QTLs) governing variability of traits involved in stress tolerance is of greatest interest if their effects can be scaled up to crop performance in the field under different stress scenarios [57]. Two strategies have been adopted for that.

The co-location of QTLs between simple traits and yield essentially applies to constitutive traits with a large effect on yield. The assumption is that the concurrent effects of each QTL on yield and the target trait are controlled by the same locus and not by tight linkage [34[•]]. In durum wheat, field testing has revealed two QTLs that consistently influence plant height, kernel weight and yield, across a broad range of soil moisture environments [7^{••}]. Hence, the fine mapping and eventually cloning of these QTLs will be facilitated by measuring plant height rather than yield. Once the cloning of these two QTLs is completed, a better understanding of the mode of action and agronomic value of the allelic series present at these loci will be possible [58].

Another strategy involves the use of simulation modelling, and can be used for adaptive traits whose correlation with yield varies greatly according to the stress scenario. In a pioneering work, Chapman *et al.* [59] have simulated the effect of the genetic variability of four traits on sorghum yield over 100 growing seasons with contrasting climatic scenarios. Afterwards, they calculated the proportion of cases in which a given allele has a positive effect on yield over a long climatic series, under a broad range of simulated cropping systems on the basis of different sowing dates, plant density or fertilizer application, as well as forecasted climate changes. Therefore, one can weigh the agronomic value of an allele in multiple combinations of sites, climatic conditions and cultivation techniques involving thousands of virtual cases, clearly an objective impossible to pursue experimentally. A recent study has extended this approach to the experimental QTLs of leaf growth sensitivity to water deficit originat-

ing from Figure 1 [19^{••}]. Through the model, the QTL allelic effects on the sensitivity of leaf elongation rate to environmental conditions were scaled up into the allelic effects on yield in a large range of environmental scenarios [19^{••},40] (Figure 2). Notably, Figure 2 shows that a positive allelic effect for leaf growth sensitivity translated into either a positive or a negative effect on yield in a given site, depending on the timing and intensity of the stress.

Conclusion

Fifteen years ago, Passioura argued whether simulation models are science, snake oil, education or engineering [60]. This is a healthy question that still deserves consideration. The dissection-modelling approach reviewed here allows progress in the understanding of the stress effects (‘science’), and potentially in the selection process (‘engineering’). It has serious limitations in terms of reliability (‘snake oil’) if it is not appropriately associated with field experiments analysed with suitable statistical methods [5^{••}]. The role of field experiments, in this approach, is to test to what extent simulations can predict yield in a subset of conditions, rather than to explore the whole range of conditions in which the agronomic value of an allele is assessed.

The approach presented herein should be considered as a contribution of Genetics and Ecophysiology to Systems Biology scaled up at the whole-plant level and aimed at bridging the gap between functional polymorphisms and tolerance to abiotic stresses in crops. Its application relies on high-throughput, accurate phenotyping combined with modelling, backed up by multi-environment statistical analyses [5^{••},21] and high-resolution genetic dissection made possible via genomics approaches [61]. Clearly, the challenge to deploy the deluge of information produced by genomics to predict crop performance under stressed field conditions is in its infancy. Modelling holds promise to streamline and speed up the tortuous gene-to-phenotype journey [6^{••}], thus providing breeders with an additional tool for tailoring cultivars able to better cope with the consequences of global climate change and suitable for a more sustainable agriculture.

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