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QTL × environment interactions in rice. I. Heading date and plant height

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Abstract One hundred twenty six doubled-haploid (DH) rice lines were evaluated in nine diverse Asian environments to reveal the genetic basis of genotype × environ-

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ment interactions (GEI) for plant height (PH) and heading date (HD). A subset of lines was also evaluated in four water-limited environments, where the environmental basis of $G \times E$ could be more precisely defined. Responses to the environments were resolved into individual QTL \times environment interactions using replicated phenotyping and the mixed linear-model approach. A total of 37 main-effect QTLs and 29 epistatic QTLs were identified. On average, these OTLs were detectable in 56% of the environments. When detected in multiple environments, the main effects of most QTLs were consistent in direction but varied considerably in magnitude across environments. Some QTLs had opposite effects in different environments, particularly in waterlimited environments, indicating that they responded to the environments differently. Inconsistent QTL detection across environments was due primarily to non- or weakexpression of the QTL, and in part to significant $QTL \times$ environment interaction effects in the opposite direction to QTL main effects, and to pronounced epistasis. QTL × environment interactions were trait- and gene-specific. The greater GEI for HD than for PH in rice were reflected by more environment-specific QTLs, greater frequency and magnitude of OTL × environment interaction effects, and more pronounced epistasis for HD than for PH. Our results demonstrated that QTL × environment interaction is an important property of many QTLs, even for highly heritable traits such as height and maturity. Information about QTL × environment interaction is essential if marker-assisted selection is to be applied to the manipulation of quantitative traits.

Keywords Genotype \times environment interactions \cdot Epistasis \cdot QTL mapping \cdot *Oryza sativa* L.

Introduction

Most plant traits are quantitative in nature and are influenced by many genes or quantitative trait loci (QTLs). Quantitative traits are also influenced by the environment and tend to show varied degrees of genotype × environment interactions (GEIs). GEIs occur when two or more genotypes perform differently in different environments, and are thus described as differential genotypic sensitivities to environments (Falconer 1981). Plants, particularly self-pollinated plants, tend to show a high level of GEIs that allow better adaptation to their changing environments and the maintenance of genetic variation in populations (Jain and Marshall 1967). In plant breeding, GEIs must be considered to identify superior and stable genotypes when breeding materials are tested in different environments. Because of their importance in plant breeding and evolution, GEIs of quantitative traits have been the subject of extensive investigations (cf. Baker 1988; Cooper and Hammer 1996). Classical studies on GEIs using segregating plant populations have been few, but they have yielded valuable information regarding the importance of GEIs for quantitative traits (Mather and Jinks 1982).

DNA markers and high-density genetic maps of major crops developed since the late 1980s have facilitated efforts to understand the genetic basis of quantitative traits through QTL mapping. Main-effect QTLs (M-QTLs) affecting a wide range of agronomic traits in many plant species have been reported (cf. Paterson 1995; Georges 1997; Stuber 1997). Despite the technical difficulties, OTL × environment interaction has been revealed by inconsistent detection and variable effects of M-QTLs across environments in tomato (Paterson et al. 1991), maize (Bubeck et al. 1993; Veldboom and Lee 1996; Austin and Lee 1998; Crossa et al. 1999; Jiang et al. 1999), barley (Hayes et al. 1993), and rice (Lu et al. 1996; Zhuang et al. 1997). In soybean, QTLs were inconsistent across environments for plant height and lodging resistance, but consistent for maturity, indicating that $QTL \times$ environment interaction is trait dependent (Lee et al. 1996).

In most previous studies, QTL × environment interaction was inferred by comparing QTLs detected in different environments. This inference about the presence of QTL × environment interaction has two shortcomings. First, individual OTL \times environment interaction effects were not properly quantified, largely because of a lack of appropriate analytical methodology. Second, because only a single threshold was used in most QTL mapping studies, it remains unknown whether inconsistent QTL detection was due to the type-II error arising from the use of single thresholds or to true differential trait expression across environments. Using composite interval mapping, Tinker et al. (1996) were able to detect considerable QTL × environment interaction for seven agronomic traits in two barley crosses, even though many of the detected QTLs were highly consistent across environments. Yan et al. (1999) reported significant QTL × environment interaction associated with common QTLs for plant-type traits in a rice doubled-haploid (DH) population in two different environments. Unfortunately, none of these studies was able to dissect GEIs in the presence of epistasis, which underlies complex phenotypes (Yu et al. 1997; Li et al. 2001, Luo et al. 2001). Thus, many important questions regarding the genetic aspects of GEIs remain largely unanswered. These include the following: Which QTLs are more environment-specific? How frequently do QTLs interact with environments? If present, how important are QTL \times environment interaction effects as compared to QTL main effects?

This manuscript describes a large study of QTL \times environment interaction in a diverse range of environments for the genetic control of two highly heritable traits, plant height (PH) and heading date (HD), in a well-known doubled-haploid (DH) rice population. The primary objective of this work is to quantify M-QTLs and epistasis with regard to their interactions with environments.

Materials and methods

The experimental population and field trials

The mapping population used in this study comprised 135 DH lines derived from the cross between an indica variety, IR64, and an upland japonica variety, Azucena, as described previously (Huang et al. 1994). Two phenotyping experiments were conducted. Experiment 1 was conducted during 1994 and 1995 in nine diverse environments including seven locations in four Asian countries (Philippines, China, India and Thailand) and two different growing seasons at two of the locations. The geography of the environments covered a wide range of latitudes and longitudes from 13.5° to 31.5° N and from 76° to 121.5° E (Table 1). The same set of DH lines and the parents were evaluated in each of the nine environments in a randomized complete block design with two or three replications. A spacing of 30×20 cm between rows and between plants within a row was used at all locations except E2 (30×25 cm) and E7 (30×15 cm); plot size was 3 to 6 rows, with 15 plants per row. All experiments were established by transplanting and maintained as lowland fields with standing water present for most of the season. The management of the field experiments was in accordance with local standard practices. Twelve traits, including plant height (PH), heading date (HD), and grain yield and its components, were measured on five representative plants in each plot. PH (in cm) was measured from the soil surface to the tip of the tallest panicle of each plant at maturity. HD was recorded as days from the time of sowing to that of the first panicle flowering in 50% of the plants in each plot. Because nine DH lines were segregating for the measured traits, apparently because of outcrossing, they were not included in subsequent data analyses. Experiment 2 was conducted with the parents plus a subset of 82 DH lines during the dry seasons of 1998 and 1999 in aerobic soil conditions under two water levels. The lines were selected to reduce the range of heading dates. These experiments were established by direct sowing at a seed rate of 80 kg ha⁻¹. Entries were replicated twice in the control section and twice in the stress section. In 1998, plots were 3-m long on beds and 0.9-m apart. Three rows were sown on each bed and the spacing between rows within a plot was 18 cm. The area received sprinkler irrigation twice each week until 30 days after planting, after which furrow irrigation was applied twice each week. This level of irrigation resulted in a fully aerobic soil profile where soil moisture status remained near field capacity. In the stress section, water was withheld from 52 to 64 days after planting and again from 70 to 84 days after planting. In 1999, plots were 3 m by 1 m, with 25 cm between rows. The crop was established using sprinkler irrigation, and then changed to drip irrigation three times a week. This type of water management applied water equivalent to about 1.4-times the potential evapotranspiration, and the soil remained aerobic. In the stress treatment, water was withheld for 14 days, beginning, on average, 12 days before HD in that plot. After

Table 1 The test environments where the IR64/Azucena doubled-haploid population was evaluated

Code	# of replications	Environments
E1	3	International Rice Research Institute (IRRI), Los Banos, Philippines, N 14.2°, E 121.5°, 1994 wet season
E2	2	IRRI, 1994–1995, dry season
E3	3	China National Rice Research Institute, Hangzhou, China, N 31.5Ű, E 120.5Ű, May–Oct, 1995
E4	2	University of Agricultural Sciences, GKVK, Bangalore, India, N 13.5Ű, E 76.5Ű, May–Oct, 1995
E5	3	South China Agricultural University, China, N 23°, E 113.5°, early season (Feb–July), 1995
E6	3	South China Agricultural University, China, N 23°, E 113.5°, late season (July–Nov), 1995
E7	2	Indian Agricultural Research Institute, New Delhi, N 28.5Ű, E 77Ű, May–Oct, 1995
E8	2	Punjab Agricultural University, Ludhiana, India, N 31.5Ű, E 76Ű, May–Oct, 1995
E9	2	Rice Research Institute, Bangkok, Thailand, N 14.5Ű, E 101Ű, May–Oct, 1995
98N	2	IRRI, 1998–1999 dry season under well-watered aerobic conditions
98D	2	IRRI, 1998–1999 dry season with water stress
99N	2	IRRI, 1999–2000 dry season under well-watered aerobic conditions
99D	2	IRRI, 1999–2000 dry season with water stress

the stress treatment, irrigation three-times per week was resumed. Data on PH and HD were collected as described in experiment 1.

Linkage map construction and data analyses

A total of 178 markers, including 147 RFLPs, 8 isozymes, 11 RAPDs and 12 cloned genes, was used to construct a complete linkage map for the DH population, as described previously (Huang et al. 1997). This map covers all 12 rice chromosomes with a total genome size of 2,003.4 cM and an average distance of 12.4 cM between adjacent markers. Analysis of variance was performed to evaluate differences among the DH lines and between the parents, among the environments, and the GEIs for the measured traits using the SAS PROC GLM (SAS Institute 1996). Correlation between the two traits in each of the environments was determined using the SAS PROC CORR (SAS Institute 1996).

Mixed linear models for QTL analyses

According to classical quantitative genetics theory (Falconer 1981), the phenotypic value of a DH line (y_{hk}) in a specific environment can be described by the following genetic model:

$$y_{hk} = \mu + E_h + G_k + GEI_{hk} + \varepsilon_{hk} , \qquad (1)$$

where y_{hk} is the phenotypic effect of the *k*th DH line in the *h*th environment, μ is the population mean, $G_k \sim (0, \sigma^2_G)$ is the genotypic effect of the *k*th DH line in the *h*th environment, $E_h \sim (0, \sigma^2_E)$ is the effect of *h*th environment, $GEI_{hk} \sim (0, \sigma^2_{GE})$ is the effect of *h*th environment, $GEI_{hk} \sim (0, \sigma^2_{GE})$ is the effect of *h*th environment, $GEI_{hk} \sim (0, \sigma^2_{eE})$ is the residual. The genotypic effects (G_k), environmental effects (E_h) and GEI effects (GEI_{hk}) can be predicted by the adjusted unbiased prediction method, in which $y_{k(G)}=\mu+G_k$ and $y_{hk(GE)}=\mu+E_h+GEI_{hk}$ (Zhu and Weir 1996; Yan et al. 1998). The mean phenotypic values, y_{hk} , of individual DH lines and their GEI effects ($GEI_{hk}=y_{hk}-y_{k(G)}-E_h+\mu$) were used as input data for identifying M-QTL and epistasis affecting PH and HD using the following mixed linear model and the corresponding computer software, QTLMAPPER v. 1.0 (Wang et al. 1999):

$$y_k = \mu + A_i x_{A_{ik}} + A_j x_{A_{jk}} + A A_{ij} x_{AA_{ijk}} + \sum_f u_{M_{fk}} e_{M_f} + \sum_l u_{MM_{lk}} e_{MM_l} + \varepsilon_k , \qquad (2)$$

where y_k is the phenotypic value of a quantitative trait measured on the *k*th DH line (*k*=1, 2,..., *n*); μ is the population mean; A_i and A_j are the additive effects (fixed) of two putative QTLs (Q_i and Q_j), respectively; AA_{ij} is the additive by additive effect (fixed) between Q_i and Q_j ; x_{Aik} , x_{Ajk} and x_{AAijk} are coefficients of QTL effects derived according to the observed genotypes of the markers (M_{i-} , M_{i+} and M_{j-}, M_{j+}) and the test positions $(rM_{i-}Q_i \text{ and } rM_{j-}Q_j)$; $e_{M_f} \sim N(0, \sigma_M^2)$ is the random effect of marker f with indicator coefficient $\mu_{M_{fk}}$ (1 for M_fM_f and -1 for m_fm_f); $e_{MM_l} \sim N(0, \sigma_{MM}^2)$ is the random effect of the *l*th marker interaction (between marker K_l and marker L_l) with indicator coefficient $\mu_{M_{lk}}$ (1 for $M_KM_KM_LM_L$ or $m_Km_Km_Lm_L$ and -1 for $M_KM_Km_Lm_L$ or $m_Km_KM_LM_L$); and $\varepsilon_k \sim N(0, \sigma_e^2)$ is the random residual effect. The inclusion of e_{M_f} and e_{MM_l} in the model is intended to absorb the additive and epistatic effects of background QTLs (additional segregating QTLs other than the loci examined) to control the noise caused by them (Wang et al. 1999).

Mapping QTLs and quantifying QTL × environment interaction for data of experiment 1 were carried out in two steps. In the first step, GEI effects (GEI_{hk}) of individual DH lines in each of the environments were obtained according to model (1) (Zhu and Weir 1996). Because the sampled environments represented a random sample of the major rice-growing environments in Asia, E_h and GEI_{hk} in model (1) were considered as random factors. Next, the mean trait values (y_{hk}) and GEI values (GEI_{hk}) of individual DH lines from each of the environments were used as input data to identify QTLs contributing to trait variation among the DH lines and QTLs contributing to the GEI variation in each of the environments using model (2) (Wang et al. 1999). Thus, M-QTLs and epistatic QTLs (E-QTLs) identified by model (2) using the mean trait values (y_{hk}) of individual DH lines as input data were expected to contain confounded QTL effects (A_i and A_{ii}) and QTL × environment interaction effects ($A_i E_h$ and $A A_{ij} E_h$), while those QTLs obtained using the predicted \overrightarrow{GEI} effects ($\overrightarrow{GEI}_{hk}$) as input data were expected to be largely due to A_iE_h and $AA_{ij}E_h$ effects. The software, QTLMAPPER v. 1.0 based on model (2) (Wang et al. 1999) first identified significant markers (M-QTLs) and marker pairs (E-QTLs) associated with the mean trait or GEI values using stepwise regression with a threshold of $P \le 0.005$ and $P \le 0.001$, respectively. Then, QTL parameters (locations, effects and test statistics) of all putative M-QTL and E-QTL pairs were estimated using the interval mapping and the restricted maximum-likelihood estimation method with all those markers identified in the first step fixed in the model to control the background genetic variation. The permutation method was used to obtain the empirical thresholds (2.97 for HD, 2.86 for HD-GEI effects, 2.87 for PH, and 2.85 for PH-GEI effects) of the experiment based on 1,000 runs of randomly shuffling the trait values, which were expected to have a genomewide type-I error of P=0.05 (Churchill and Doerge 1994). Thus, the LOD=2.9 was used for claiming significant M-QTLs in this study. For E-QTLs, a threshold of LOD=3.0 was used. We realized that the use of a single arbitrary threshold in QTL mapping could easily detect a QTL in one environment but not in another. Thus, while all QTLs detected at the selected thresholds are presented, any QTL detected in only 1 or 2 environments was interpreted with caution. Also, to examine the extent of type-II error causing inconsistent QTL detection across the environments, all identified M-QTLs and E-QTLs were re-examined in the environments using the mean trait values and trait GEI values of individual DH lines from specific environments under the minimum threshold of P<0.05. In other words, when a QTL was identified using the mean trait or GEI values in one environment, this QTL was also tested by the data from all other environments, and the test statistics and QTL parameters associated with the QTL are also reported as long as the QTL reached the minimum threshold. For experiment 2, only M-QTLs were examined using the mean trait values of individual DH lines because of the small population size.

Results

Phenotypic variation of the DH lines

Table 2 shows summary statistics of the phenotypic performance of the DH lines and parents for PH and HD in the nine environments. Azucena was much taller than IR64 in all environments, but the difference in height between the parents varied considerably across the environments, ranging from 34.1 cm in E1 to 66.0 cm in E2. On average, the parents had a similar HD (102.4 days for Azucena and 103.2 days for IR64), but significant differences in HD between the parents were detected in all environments except E4. Azucena headed earlier than IR64 in E3, E5 and E7, but later than IR64 in the other environments. In experiment 2, water stress was severe in 1998 and caused reduced height and delayed heading in the parents (26.6 cm and 6 days for Azucena, and 22.6 cm and 11 days for IR64). In 1999, the stress was milder, but still caused 11-days heading delay and 8.9 cm of height reduction for Azucena, and 6-days heading delay and 24.1 cm of height reduction for IR64. Similar to the parents, the DH lines, on average, had 11.8days delayed heading and 16 cm of reduced height in 1998 and only 1.7-days heading delay and 1.4 cm of height reduction in 1999.

The DH lines showed transgressive segregation for both traits in all environments (Table 2) and showed much greater GEIs than the parents. ANOVA indicated that variances among the genotypes (the parents and DH lines), the nine environments, and GEIs were highly significant for both PH and HD. However, the relative contributions of the variance components to the total variation were different for the two traits. HD was influenced more by the environments and showed a much greater GEI than PH. For PH, the variances among the genotypes, among the environments, and the GEIs accounted for 63%, 16% and 12% of the total phenotypic variation, respectively. For HD, these components explained 18%, 54% and 20% of the total phenotypic variation, respectively. In experiment 2, ANOVA indicated that for PH, the variances among the DH lines $(R^2=70\%)$, between the years $(R^2=5\%)$ and between the water levels ($\mathbb{R}^2=8\%$), and genotype \times water \times year $(R^2=4\%)$ were significant except for genotype × water. For HD, only the variances among the DH lines $(R^2=27\%)$ and between different water levels $(R^2=56\%)$ were statistically significant.

In experiment 1, the mean height of the DH lines varied greatly across the environments, indicating the large environmental effects on height. The DH lines and parents had a below-average PH in E1, E2, E3 and E4, but were taller in E5, E7, E8 and E9. Similarly, heading tended to be earlier in E1, E2 and E8, whereas heading was delayed for both DH lines and the parents in E5, E7 and E9. There was no obvious correspondence between delayed heading and short daylength in the environments, indicating that the parents and DH lines were not photoperiod-sensitive. The coefficients of variation of the DH population for both traits remained largely consistent across the environments, except that the DH lines showed a much greater variation for HD in E1. There was a low but significant positive correlation between PH and HD in the DH lines (r=0.24, 0.23, 0.17,0.48, 0.31 and 0.23 in E1, E2, E3, E4, E6 and E9, respectively), but correlation was not significant in E5, E7 and E8. In experiment 2, the trait variation in the DH population remained unchanged across years and water-

 Table 2
 Summary statistics of phenotypic performance of the IR64/Azucena doubled-haploid (DH) population and its parents for plant height (cm) and heading date (days)

	Heading da	ate				Plant height					
_	Parents		DH population			Parents		DH population			
Env. E1 E2 E3 E4 E5 E6 E7 E8 E9	Azucena 80.0 102.0 98.7 105.0 116.4 104.6 104.5 102.5 108.0	IR64 75.0 93.0 107.3 106.5 123.3 102.4 118.5 97.0 106.0	$\begin{array}{c} \text{Mean } \hat{A} \pm \text{SD} \\ 90.3 \hat{A} \pm 10.7 \\ 99.3 \hat{A} \pm 6.8 \\ 110.3 \hat{A} \pm 7.3 \\ 104.1 \hat{A} \pm 5.7 \\ 122.9 \hat{A} \pm 7.1 \\ 106.6 \hat{A} \pm 8.3 \\ 111.2 \hat{A} \pm 5.4 \\ 102.2 \hat{A} \pm 5.0 \\ 109.6 \hat{A} \pm 6.4 \end{array}$	Range 61.0-110.0 84.0-120.0 89.7-124.5 83.5-115.0 101.1-142.1 80.9-125.1 100.1-124.5 84.0-113.0 94.7-129.0	CV% 11.8 6.9 6.6 5.4 5.7 7.8 4.9 4.9 5.9	Azucena 123.0 142.0 127.0 143.1 153.8 149.5 158.8 159.3 167.2	IR64 88.9 74.0 83.8 86.7 101.5 103.7 107.8 105.7 117.4	$\begin{array}{c} \mbox{Mean} \ \hat{A} \pm \ SD \\ 96.9 \ \hat{A} \pm 16.3 \\ 102.3 \ \hat{A} \pm 20.3 \\ 100.8 \ \hat{A} \pm 16.2 \\ 105.5 \ \hat{A} \pm 21.8 \\ 120.1 \ \hat{A} \pm 23.2 \\ 112.9 \ \hat{A} \pm 20.1 \\ 119.9 \ \hat{A} \pm 21.8 \\ 115.1 \ \hat{A} \pm 21.9 \\ 125.9 \ \hat{A} \pm 23.0 \end{array}$	Range 59.3–139.7 58.8–157.0 66.7–136.5 61.7–157.9 73.4–181.4 74.2–160.0 76.7–178.0 69.5–176.5 70.9–183.9	CV% 16.8 19.9 16.1 20.7 19.3 17.8 18.2 19.0 18.2	
98N 98D 99N 99D	94.0 100.0 90.0 101.0	85.0 96.0 87.0 93.0	92.0ű9.3 103.8ű10.5 94.7ű8.6 96.4ű8.6	70.0–113.0 73.0–121.4 69.5–112.0 69.0–111.0	10.1 10.1 9.1 8.9	116.3 89.7 133.1 124.2	82.8 60.2 93.0 68.9	94.7ű12.8 78.7ű10.4 87.3ű13.7 85.9ű14.3	69.5–112.0 57.5–100.1 56.0–122.0 58.7–122.5	13.5 13.2 15.7 16.6	

^a SD and CV are the standard deviation and the coefficient of variation

M-QTL	Chr.	Marker interval	Para.	E1	E2	E3	E4	E5	E6	E7	E8	E9	98N	98D	99N	99D
sd-1	1	RZ730– RG810	LODa Effect AE	18.7 -9.3 2.5***	13.5 -12.9	17.2 -8.2 6.1***	23.7 -16.0 -1.8**	25.2 -14.7 -1.2**	16.0 -14.4	16.0 -9.9	18.0 -13.4	24.0 -18.3 -2.8***	16.3 -8.7	9.9 -7.5	12.6 -9.3	9.5 -8.0
QPh1	1	RG345– RG381	LOD Effect AE		1.8 2.8	-2.4***	1.7 -2.6	2.6 -2.8	2.3 4.7	4.6 4.4 -1.2***	3.6 6.7 1.6*	3.8 6.1 -1.5**				2.4 -2.9
QPh2a	2	RG654– RG256	LOD Effect AE	2.5 2.1 -3.4***	2.7 3.6 1.3**	2.8 2.8	9.2 5.9	6.9 6.4 2.7***		9.0 4.7	5.9 6.3					
QPh2b	2	RG157– RZ318	LOD Effect					2.5 2.8			3.7 3.7					
QPh3a	3	RG348– RZ329	LOD Effect AE	2.6 -2.7	3.1 -3.2 -2.0***											1.7 1.6
QPh3b	3	RZ394– RZ284	LOD Effect AE	-2.7***	3.9 3.3	2.5 -2.2* -2.2*	2.5 3.3	3.0 4.4 1.6*	3.9 5.4	6.5 6.4 2.8***	4.8 4.4	4.7 5.4 2.7***			1.8 1.8	2.1 2.4
QPh3c	3	CDO87– RG418a	LOD Effect	13.4 -7.0	9.4 -6.6	6.6 -6.7	9.4 -7.7	6.4 -7.9		8.3 -6.3	5.3 -5.6	4.0 -3.7	2.5 -5.7		3.8 -5.6	3.6 -5.0
QPh4a	4	RG908– RG190	LOD Effect AE	3.6 3.1	4.0 4.5 -2.1**	4.8 4.6 -2.2**	5.3 4.4	5.5 4.7 2.8***	4.8 5.3	8.7 7.1 2.9***	-2.4**	8.6 7.5	3.2 -3.8	2.0 -2.8		
QPh4b	4	RG449– RG788	LOD Effect AE			3.2***	4.6 -4.4		1.1*	-2.0**		1.1*	6.8 -4.6			
QPh4c	4	RZ590– RG143	LOD Effect AE	4.3 -3.9 -2.4***	5.9 -4.4	4.2 -4.0	3.8 -5.4 -2.0***	10.3 -8.0 -2.9***		3.7 -4.2	7.3 -6.8	2.6 -3.5	2.4 -3.0	1.8 -1.9		
QPh5	5	CDO105– RZ649	LOD Effect AE	3.1 5.2	2.0 2.9	3.0 4.5 -2.8*	2.4 3.1	1.9 2.3	3.7 6.6	3.0 3.2	4.6 8.5 3.9***	3.8 5.2	2.9 4.6			1.8 -2.2
QPh7	7	RZ488– RG477	LOD Effect AE	2.4 2.3 -1.6**	-3.0***		1.9***			3.5 4.8 1.8**						2.6 4.7
QPh8	8	TGMS1.2– AG8	LOD Effect AE	2.5 2.8	5.1 3.8 2.6***		1.9 2.7 0.8*	-2.0**		4.8 2.2 -1.1*	-2.2**		2.3 -3.1	2.7 -3.6	6.1 -7.1	4.9 -6.8
QPh9a	9	RZ206– RZ422	LOD Effect AE	2.1 -3.3	1.9 -2.5	-3.9	2.1 -3.3	3.8 -5.9 -1.6**	-1.9***	2.1 -1.8 1.3**	1.9 -3.1	3.1 -5.0	3.5 -3.3	2.0 2.2	2.1 -2.2	2.8 3.2
QPh9b	9	Amy3ABC– RZ228	LOD Effect							6.2 -4.9						
QPh11	11	RG247– RG167	LOD Effect AE			2.1 -3.1 -2.6***						4.9 2.5 2.6***			1.9 -2.6	
QPh12	12	AF6– RG457	LOD Effect	3.6 3.6		3.4 3.4									2.6 3.0	3.3 3.3

Table 3 Main-effect QTL for plant height and their interactions with environments detected in the IR64/Azucena doubled-haploid population

Bold parameters were detected in epistatic models. *, **, *** represent the significance levels of t tests at $P \le 0.05$, $P \le 0.01$, and $P \le 0.001$, respectively

^a Effect and AE were QTL phenotypic effect and QTL by environment interaction effects, the former was estimated from mean trait values of individual DH lines in individual environments and the latter estimated from GEI effects of individual DH lines in individual environments. The sign indicates the direction of the effect of the IR64 allele

stress conditions, but PH showed slightly reduced variation and HD exhibited increased variation as compared to experiment 1 (Table 2). Significant positive correlation (r=0.34) between the two traits was detected only in 1998 under stress.

M-QTLs and E-QTLs for PH and their interactions with environments

In experiment 1, QTL mapping based on the data of individual environments led to the identification of 17 M-QTLs on all 12 rice chromosomes except chromosomes 6

and 10 (Table 3, Fig. 1). The average number of detectable M-QTLs per environment was 9.8 ± 2.0 , ranging from 5 in E6 to 12 in E7. The M-QTLs with the largest effect mapped between RZ730 and RG810 on chromosome 1 in the same region as the semidwarf gene, *sd-1*. Here, the IR64 allele reduced height by an average of 13.0 ± 3.3 cm (ranging from 8.2 to 18.3 cm). Detected in all 13 environments, this QTL showed significant interactions with E1, E3, E4, E5 and E9, with significant additive × environment (*AE*) effects ranging from -2.8 to 6.1 cm. Other important M-QTLs for PH included *QPh5* detectable in all nine lowland environments; *QPh3b*, *QPh3c*, *QPh4a*, *QPh4c* and *QPh9a* detected in eight of



Fig. 1 Genomic locations of identified main-effect QTLs affecting plant height (PH) and heading date (HD) of the IR64 \times Azucena DH population, detected in the nine environments of experiment 1.

Ph2, *Ph3*, etc. were M-QTLs affecting PH detected in the same population reported by Yan et al. (1998)

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Table 4 Epistatic QTL pairs affecting plant height and their interactions with environments in the IR64/Azucena DH population

Chr.	Marker <i>i</i>	Ch.	Marker j	Para.	E1	E2	E3	E4	E5	E6	E7	E8	E9	Mean ± SD
1	RG381	3	CDO337	$\begin{array}{c} \text{LOD}^{\text{a}} \\ \text{Effect}^{\text{b}} \\ AA_{ij}E^{\text{b}} \end{array}$	3.1***	1.8 3.7 * 2.3**	4.10 4.4 2.4**	1.7 2.6				-3.4***	k	3.57±0.91 1.10±3.02
1	RZ276	5	RZ67	LOD Effect	5.1 8.5	2.5 3.6	5.0 5.9	4.7 7.7	5.0 5.6	3.7 7.2	3.0 4.3	8.5 12.5		6.91±2.80
1	RG246	10	RG257	LOD Effect AA _{ij} E	3.0 3.4	2.2 2.2 -1.3*	-3.1***	3.2 4.7	3.0 4.3 1.7**	4.0 5.9	4.4 4.6	1.8 4.3	3.6 6.5 1.9***	4.49±1.34 -0.20±2.42
2	RG544	9	G103	LOD Effect		3.2 4.7	1.8 2.4	2.3 3.9	2.4 3.5			1.9 3.2		3.54±0.85
3	RZ284	5	RG13	LOD Effect	4.5 5.0	3.9 4.3		2.5 3.7	3.0 4.3	3.9 4.6		4.8 5.0	2.7 3.3	4.31±0.64
3	RZ284	11	RG103	LOD Effect AA _{ij} E	2.0 2.8	3.2 4.8	-1.4*				-1.3*	2.0 1.8	2.3 3.1	3.13±1.25 -1.35±0.07
4	RG190	6	RG424	LOD Effect AA _{ij} E	2.2 3.7			4.0 4.8	4.1 3.4	3.3 5.1	3.1 5.0 1.6**	-1.7**	4.7 4.8 1.7**	4.47±0.73 0.53±1.93
4	RG163	8	RZ66	LOD Effect AA _{ij} E	2.6 -3.8 1.6**	5.2 -6.1	5.0 -5.4	6.0 -4.6 -0.8*	3.7 -5.8 -1.6**	4.3 -4.8 1.6**	3.8 -5.7	3.5 -4.1	4.8 -6.1 -2.2**	-5.16±0.86 -0.28±1.79
5	RG13	11	RZ536	LOD Effect	5.0 4.7	4.0 4.6	3.1 3.3	3.4 5.7	3.3 5.6	3.3 6.2		6.9 8.4	4.0 7.0	5.69±1.57
9	RZ206	11	RG1109	LOD Effect	4.2 4.6	2.4 3.6	5.0 5.8	4.2 4.6	8.2 5.7	5.1 6.5	2.9 3.0	3.0 4.1	5.0 5.4	4.81±1.14

*, **, and *** represent the significance levels of $P \le 0.05$, 0.01 and 0.001, respectively based on t tests

^a DNA markers more closely linked to the corresponding E-QTLs and the bold markers represent main-effect QTLs

^b Effect is the phenotypic epistatic effect estimated from the mean trait values of individual DH lines, while $AA_{ij}E$ is the epistasis by environment interaction effect estimated from GEI effects of individual DH lines in each of the environments. The sign indicates the direction of the parental-type interaction effects associated with the two parental digenotypes, I^{IR64}J^{IR64}J^{IR64}J^{IR64} and I^{AZU}J^{AZU}

the lowland environments; and *QPh1* and *QPh2a* in seven of the environments. Of the M-QTLs mentioned above, we noted that QPh1, QPh3b, QPh5 and QPh9a showed large variation in their effects (in both magnitude and direction). The remaining M-QTLs appeared to be more or less environment-specific. These included QPh8 detected in four of the lowland environments; QPh2b, QPh3a, QPh7, QPh11 and QPh12 in two of the lowland environments; and QPh4b and QPh9b in only one lowland environment. Significant AE effects were detected in 46 cases (30%) for all M-QTLs, except for Ph2b, QPh3c and QPh9b. The IR64 allele at QPh2a, QPh2b, QPh4a, QPh5, QPh7 and QPh12 increased height in lowland environments, while it reduced height at the remaining M-QTLs except for QPh1, QPh3b and QPh11, at which QTL main effects differed in both magnitude and direction across the lowland environments.

In experiment 2, 14 of the 17 PH M-QTLs detected in experiment 1 were also identified in one or more cases except for *QPh2a*, *QPh2b* and *QPh9b* (Table 3). Aerobic soil conditions resulted in a reverse in the effect of the IR64 allele at *QPh3a*, *QPh4a* and *QPh8* compared to the flooded experiment 1. Water stress caused further differences in the detected M-QTLs. Drought reduced the phenotypic effects of *sd-1*, *QPh3c*, *QPh4a* and *QPh4a*, and *QPh4b*, increased the effects of *QPh3b*, *QPh8* and *QPh12*, and

switched the direction of the QTL effects relative to the well-watered aerobic control and the flooded trials in experiment 1 for both *QPh5* and *QPh9a*.

In addition to the M-QTLs mentioned above, ten E-QTL pairs were identified for PH in experiment 1. Of these, five interactions occurred between an M-QTL and a modifying factor (Table 4). On average, each of the E-QTLs was detectable in 6.7 (74%) of the environments. Of these, two E-QTL pairs were identified in all nine environments, two in eight of the environments, one in seven of the environments, one in six of the environments, one in five of the environments, one in four of the environments and two in three of the environments. Interestingly, the AA effects of the E-QTL were consistent in direction, varying only in magnitude across the environments. Significant $AA_{ii}E$ effects were detected for six of the 11 E-QTL pairs and these $AA_{ii}E$ effects differed greatly in both direction and magnitude across the environments.

M-QTLs and E-QTLs for HD and their interactions with environments

Twenty M-QTLs affecting HD were identified in experiment 1 and they were mapped to all 12 rice chromosomes

QTLs	Chr.	Marker Interval	Para.	E1	E2	E3	E4	E5	E6	E7	E8	E9	98N	98D	99N	99D
QHd1a	1	RG472– RG246	LOD Effect		2.8 1.7		2.0 -0.9	1.7 -0.9	3.2 -2.1	2.3 -1.0		6.2 -2.2				2.1 -1.6
QHd1b	1	RG345– RG381	LOD Effect		1.1*			-0.8*		-1.2*	-1.6***	-1.5**		-3.4**		
QHd1c	1	RZ801– RG331	LOD Effect AE		2.0 -1.1	-2.1**	7.5 -2.2	5.3 -2.3 -1.2***	8.9 -3.3 -1.5**	-0.7*	1.5***	2.7 -1.3	3.4 -2.8	3.2 -3.6	1.8 -1.9	
QHd2a	2	RG171– RG157	LOD Effect AE				0.9**	1.0*	-0.7*	0.9**				-4.0**	2.6 3.3	
QHd2b	2	RZ123– RZ213	LOD Effect AE	2.8 -2.6 -1.7***	4.0 -1.6	2.3 -2.0 0.9*	1.3 -0.8	1.8 -1.3		5.9 -2.2	1.6 -1.1	1.3 0.9		2.6 2.5		
QHd3a	3	RG104– RG348	LOD Effect AE	6.4 -4.0	9.9 -3.1 -3.1***	9.2 -3.3 -1.6***	3.3 -1.9	2.0 -1.5 0.8**	5.2 -3.2	7.7 -2.1	1.6***	3.9 -1.1 1.3***	7.2 -5.3	7.6 -5.1	5.6 -5.4	5.2 -5.8
QHd3b	3	RZ678– RZ574	LOD Effect AE	2.5 2.5	-1.6***			6.0 2.2 0.8*	2.2 1.9 -0.8*	2.3 2.2		1.6 0.9		-1.3		
QHd3c	3	RZ448– Pgil	LOD Effect AE	1.4 1.1	2.4 1.8 -0.8*			8.8 3.0 1.9***	2.8 3.1	2.1 1.9	2.2 1.9	4.5 2.5	3.5 -2.6	1.9 -1.9	2.8 3.1	1.8 1.9
QHd4	4	RG908– RG190	LOD Effect AE	3.0 2.7			2.8 1.0	1.5 0.9		2.5 1.0	-1.1***	2.9 1.6		2.3 1.3		
QHd5a	5	RG556– RZ556	LOD Effect	2.7 2.3	3.5 1.7	1.8 1.3						2.7 1.4				
QHd5b	5	RZ70– RZ225	LOD Effect AE		2.1***	1.1* -1.1*		-1.0**		0.6*		5.4 -2.1	2.6 -2.2* 7.1	4.5 -6.0** 5.9		
QHd6a	6	RG213– <i>Amp3</i>	LOD Effect AE	2.3 -2.1 -1.6**						4.0 -1.3	2.6 -1.3		4.2 -4.4	3.0 -3.8		
QHd6b	6	CDO544– RG653	LOD Effect AE	2.9 2.2 1.4**	1.1* 0.7*	1.5**		1.0*					1.8**		1.8 2.0	
QHd7	7	RZ488– RG477	LOD Effect AE	4.8 3.6	5.1 1.5 -0.9*	-4.3***	7.7 2.4 -0.9*	12.2 4.1 1.3***	8.4 3.1	5.7 2.2	5.1 1.8	10.4 3.4 0.7*	1.8 2.1	2.5 2.2	3.0 3.0	1.9 2.2
QHd8a	8	RG978– RG1	LOD Effect AE	1.7 2.0	1.4**	1.9 -2.0	2.3 1.1	2.6 1.9 1.5***	1.8 1.8 -0.8*	1.2*	4.9 2.1	2.0 -1.5 1.2*		4.3 6.0	2.4 2.6	
QHd8b	8	<i>Amp2–</i> CDO99	LOD Effect	2.0 -1.2				2.2 -1.2			1.4 0.8		6.2 5.1	7.3 4.3		3.1 3.4
QHd9	9	RZ206– RZ422	LOD Effect AE	3.4 -2.6	-1.0*	1.2**	-1.5**	-1.9**	1.4 -1.5		4.5 -1.8 1.2***	-1.6**	8.1 -3.7	5.6 -4.0	2.6 -2.6	
QHd10	10	R2447– RG241	LOD Effect AE					-2.9***		12.4 -1.8				7.8 -4.5		1.8 -2.1
QHd11	11	G44– RG247	LOD <i>Effect</i>				5.8 -2.0			2.0 -1.5						
QHd12	12	AF6– RG457	LOD Effect AE	4.9 3.2 2.9***	1.3 0.9			-1.1***	2.2 3.4 1.5**	1.5 -1.0 -1.3***	1.9 1.5		1.7 1.7	2.4 -2.5		

Table 5 Main-effect QTLs for heading date and their interactions with environments detected in the IR64/Azucena doubled-haploid population

Bold parameters were detected in epistatic models. *, **, *** represent the significance levels of t tests at $P \le 0.05$, $P \le 0.01$, and $P \le 0.001$, respectively

^a Effect and AE were QTL phenotypic effects and QTL by environment interaction effects, the former was estimated from mean trait values of individual DH lines in individual environments and the latter estimated from GEI effects of individual DH lines in individual environments. The sign indicates the direction of the effect of the IR64 allele

Table 6 Epistatic QTL pairs affecting heading date and their interactions with environments detected in the IR64/Azucena DH population

Chr.	Marker i	Chr.	Marker j	Para.	E1	E2	E3	E4	E5	E6	E7	E8	E9	Mean ± SD
1	U10	3	RG348	LOD Effect	2.6 -1.6	7.5 -0.9	4.7 -0.8	3.8 -1.2		2.3 -1.3	6.6 -1.6		7.1 -1.2	-1.23±0.31
1	RG173	6	RG433	LOD Effect AA _{ij} E				3.7 2.3	3.9 1.9	1.1*	-1.2*			2.10±0.28 -0.05±1.63
1	RG690	9	RZ206	LOD Effect AA _{ij} E	2.7 2.8 2.0**	5.9 -1.8 -0.7*		3.5 -1.6			-0.8*		2.0 0.8	0.05±2.18 0.17±1.59
1	RZ801	12	RG958	LOD Effect AA _{ij} E		4.3 1.6 1.1**	2.9 1.8		5.2 1.4		2.7 1.1			1.48±0.30 1.10
2	RG544	3	RZ574	LOD Effect AA _{ij} E				4.1 -1.4 -1.1**		2.8 -1.2	4.6 0.8	6.8 -1.7 -1.3**		-0.88±1.14 -1.20±0.14
2	RG437	4	RG908	LOD Effect AA _{ij} E	2.0**		1.0*	5.4 -1.5 -1.1**		2.1 -2.0 -1.4**				-1.75±0.35 0.13±1.64
2	RZ58	12	Sdh1	LOD Effect AA _{ij} E		4.5 -1.5 -1.5***		2.4 -0.8			2.6 -1.0			-1.10±0.36 -1.50
2	RZ318	7	CDO38	LOD Effect AA _{ij} E			3.4 2.1	2.0 -0.7 -0.9*		-1.3**	-0.8*			0.70±1.98 -1.00±0.26
3	RZ337A	7	RZ337B	LOD Effect AA _{ij} E				-0.7*	4.9 1.9	2.1 2.0			2.2 0.7	1.53±0.72 -0.70
3	Pgi1	8	RZ66	LOD Effect AA _{ij} E	2.4 -3.1	2.7 -0.9 1.2*	2.3 -2.4 -1.5*	2.1 -1.6 1.1*	9.7 -1.9	5.5 -4.0 -1.9**	4.1 -1.8	4.4 -1.8	7.1 -1.7	-2.13±0.92 -0.28±1.65
4	RZ262	7	RG773	LOD Effect AA _{ij} E	3.1 2.8	5.7 2.2	2.3 1.4						-1.3**	2.13±0.70 -1.30±1.3
5	RG207	8	RZ143	LOD Effect				4.8 1.3				3.1 1.0	3.2 1.3	1.20±0.17
5	RG13	9	RZ228	LOD Effect AA _{ij} E	6.0 -3.9 -2.9***		1.0*	0.7*	2.7 -1.1		0.9**	1.9 1.1 1.4***	4.2 -1.1	-1.25±2.05 0.22±1.76
6	RG172	7	RG769	LOD Effect			2.2 -1.2		2.3 -1.9				4.8 -1.8	-1.63±0.38
7	RG477	11	G186	LOD Effect AA _{ij} E	4.1 -1.4	1.1*	1.8**	1.3***	6.9 -1.6 -1.3***	1.6***	7.3 -1.4	3.4 -0.8	8.9 -1.4 -0.9*	-1.32±0.30 0.60±1.34
7	RG477	11	RZ638	LOD Effect AA _{ij} E				5.7 1.2	7.3 1.3	7.3 3.0 2.0***		-1.0**	7.7 0.9	1.60±0.95 0.50±2.12
8	RG1	11	RZ536	LOD Effect AA _{ij} E		3.2 -1.6 -1.4**	1.8**		2.0 -1.6		5.7 -3.0 -1.3**			-2.07±0.81 -0.30±1.82
9	RZ206	9	RG667	LOD Effect AA _{ij} E	2.5 -1.6	2.5 -1.3 -1.6**		4.8 -1.6	2.5 -1.8				5.9 -1.8	-1.62±0.20 -1.60
9	RZ206	10	RZ625	LOD Effect AA _{ij} E			4.1 2.8 2.5***	-0.6*	2.3 1.1					1.95±1.20 0.95±2.19
9	RZ206	11	RG118	LOD Effect AA _{ij} E	4.3 -2.8 -2.7***	2.3 -0.8 -0.9*		1.8 1.2	2.3 1.5 0.9*		0.7*		3.8 1.6	0.14±1.91 -0.50±1.67

*, **, and *** represent the significance levels of $P \le 0.05$, 0.01 and 0.001, respectively based on *t* tests ^a DNA markers more closely linked to the corresponding E-QTLs and the bold markers represent main-effect QTLs ^b Effect is the phenotypic epistatic effect estimated from the mean trait values of individual DH lines, while $AA_{ij}E$ is the epistasis by environment interaction effect estimated from GEI effects of individual DH lines in each of the environments. The sign indicates the direction of the parental-type interaction effects associated with the two parental digenotypes, I^{IR64}J^{IR64}J^{IR64}J^{IR64} and I^{AZU}J^AZUJ (Table 5, Fig. 1). The average number of detectable HD M-QTLs per environment was 11.3±2.5, ranging from 7 in E3 to 15 in E7. Of these, QHd8a was the only one detected in all nine environments. However, this QTL showed strong interactions with environments. The IR64 allele at this QTL delayed heading in E3 and E9 but caused early heading in the remaining environments. Significant AE effects were detected between *QHd8a* and E5, E6 and E9. Other important M-QTLs included QHd2b, QHd3a and QHd7 detected in eight of the environments, QHd1c and QHd3c in seven of the environments, QHd1a and QHd9 in six of the environments, and QHd1b, QHd3b, QHd4 and QHd12 in five of the environments. The remaining M-QTLs appeared to be more environment-specific. These included *QHd5a* and QHd6b detected in four of the environments, QHd2a, *OHd6a* and *OHd8b* in three of the environments, *OHd11* in two of the environments, and OHd5b and OHd10 in only one of the environments. Four M-QTLs (QHd1b, QHd2a, QHd6b, QHd8 and QHd9) were unique in that their main effects became detectable only when each one was involved in epistasis with another locus in certain environments (Table 6). Significant additive \times environment (AE) effects were detected in 39 cases (22%) for all M-QTLs except QHd5 and QHd11. The phenotypic effects of most M-QTLs varied considerably in magnitude, and five M-QTLs (QHd1a, QHd2b, QHd8, QHd11 and QHd12) showed varied effects in both magnitude and direction across the environments.

In experiment 2, 18 of the 20 M-QTLs identified in experiment 1 were also detected in one or more cases except for *QHd5a* and *QHd11* (Table 5). Drought appeared to induce the expression of *QHd1a*, *QHd2b*, *QHd3b*, *QHd4* and *QHd10*. *QHd3c*, *QHd6a* and *QHd8b* had reduced effects under drought, while *QHd1c*, *QHd5b* and *QHd8a* had increased effects under drought. For *QHd3c*, the IR64 allele was associated with delayed heading under the high-density, furrow irrigated treatment in 1998. *QHd8b* also appeared to have a contrasting effect in lowland and aerobic environments.

Table 6 shows 19 E-QTL pairs affecting HD in experiment 1. On average, the number of detectable E-QTL pairs in each of the environments was 8.6 ± 1.9 , ranging from 5 in E8 to 11 in E5 and E6. Of these, only one E-QTL pair was identified in all nine environments, one in seven of the environments, one in five of the environments, six in four of the environments, six in three of the environments, and four in two of the environments. Significant $AA_{ij}E$ effects were detected for 12 of the 19 E-QTL pairs and the $AA_{ii}E$ effects of individual E-QTL pairs identified in multiple environments differed greatly in both direction and magnitude across the environments. We noted that 12 of the interactions occurred between an M-QTL and a modifying factor. Detected as a M-QTL in three flooded environments and three aerobic environments, OHd9 (RZ206) interacted simultaneously with four other loci on chromosomes 1, 9, 10 and 11, suggesting its regulatory function in determining HD in rice.

Discussion

In this study, the genetics of two highly heritable traits of rice were dissected into QTL main effects and their interactions with environments. We included epistasis in the linear model so that the interactions of both M-QTLs and E-QTLs with environments could be characterized. It should be pointed out that, according to our models, the OTL effects (Tables 3 and 5) estimated using trait mean values from individual environments were expected to be the QTL phenotypic effects (A+AE or AA+AAE) and those detected by GEI effects were largely attributable to AE or AAE effects. This decomposition allowed us to quantify $QTL \times environment$ interaction more accurately. The number of QTLs detected in this study was much more than any previous single-environment QTL mapping studies in rice and other species. Furthermore, we found that most (73%) M-OTLs identified in this study fell within the vicinity of the M-QTLs affecting the same traits identified in the same or different rice mapping populations reported previously (Courtois et al. 1995; Li et al. 1995; Xiao et al. 1995, 1996; Huang et al. 1996; Yano et al. 1997; Zhuang et al. 1997; Yan et al. 1998; Zhou et al. 2001), indicating that these QTLs covered a significant portion of the loci affecting PH and HD in rice at which allelic differences exist. In this study, all QTLs but one (OPh9b) were detected in more than one environment.

The pattern, frequency, magnitude, and causes of QTL × environment interaction

In this study, the identified M-QTLs and E-QTLs were, on average, undetectable in 64% (58% for M-QTLs and 74% for E-QTLs) of the environments for PH and in 51% (57% for M-QTLs and 45% for E-QTLs) of the cases for HD. As the genetic basis of GEIs, $QTL \times$ environment interaction presumably arises from differential gene expression in different environments and may conceptually occur in any of the following three cases: (1) a QTL expresses in one environment but not in another, as reflected by inconsistent detection of the QTL across environments; (2) a QTL expresses strongly in one environment but weakly in another, as indicated by the variation in its effects across environments; and (3) a QTL expresses very differently and has opposite effects in different environments. All three cases were observed in this study and the pattern of the observed differential QTL expression in this study appeared to be very complex.

Non-expression of QTLs appeared to be the primary cause of the undetectable QTLs in certain environments. In these cases (37% of M-QTLs and 30% of E-QTLs for PH, and 31% of M-QTLs and 50% of E-QTLs for HD), QTLs were undetectable even under the minimum threshold of $P \le 0.05$. On the other hand, weak QTL expression in certain environments could have resulted in a high probability of type-II error in this study if the identified QTL had not been re-examined under the

Table 7 Magnitudes of QTL main effects (A and AA) and QTL x environment interaction effects (AE and AAE) associated with plant height and heading date detected in the IR64/Azucena DH population

	Main-	effect QTI				Epistatic QTL							
	A				AE				AAE				
	N ^a	n ^b	Mean ± SD	n ^b	Mean ± SD	N^{a}	n ^b	Mean ± SD	n ^b	Mean ± SD			
PH HD	17 20	88 102	5.34±3.20 1.80±0.79	46 39	2.25±0.90 1.42±0.83	10 19	67 77	4.86±1.64 1.62±0.71	25 45	1.81±0.70 1.32±0.72			

^a N is the total number of detected M-QTL or E-QTL, and ^b n is the total number of significant QTL parameters (effects) of the M-QTL or E-QTL detected across the nine environments in experiment 1

minimum threshold. In this study, the average number of detectable M-QTLs per environment was 9.8 for PH and 11.3 for HD. However, with the selected threshold of LOD=2.9, this number became only 6.9 for PH and 5.5 for HD. Similarly, with the threshold of LOD=3.0, the average number of E-QTLs detectable per environment was 4.0 for PH and 5.0 for HD, respectively, which became 6.9 for PH and 8.2 for HD under the minimum threshold (Tables 4 and 6). In other words, on average, 4.4 (41.2%) of the M-QTLs and 3.1 (40.4%) of the E-QTLs per environment would have gone undetected because of the type-II error of the selected single thresholds arising from weak QTL expression in certain environments. The third cause of inconsistent QTL detection across environments was the significant QTL × environment interaction effects. In this study, statistically significant AE effects were detected in 26% (30% for PH and 22% for HD) of the QTLs by environment combinations for the M-QTLs, and in 27% of the cases (28% for PH and 26% for HD) for the E-QTLs, as summarized in Table 7. In contrast to the QTL phenotypic effects, which were largely consistent in direction, individual AE and AAE effects varied considerably in both magnitude and direction, and appeared to be responsible for 8.2% (6.8% for M-QTLs and 9.7% for E-QTLs) of the undetected QTLs in certain environments. In these cases, the estimated AE or AAE effects were in the opposite direction to the QTL main effects. The average magnitude of the AE effects was 2.25 cm for PH and 1.42 days for HD, or 42% and 79% of the mean QTL main effects estimated in single environments, respectively. The average magnitude of AAE was 1.89 cm for HD and 1.40 days for HD, or 37% and 81% of the mean QTL epistatic effects estimated in single environments, respectively (Table 7). This was not surprising since AE or AAE effects were part of the QTL main (phenotypic) effects according to model (1).

In seven cases (*QPh1*, *QPh3b*, *QPh11*, *QHd1a*, *QHd2*, *QHd8* and *QHd12*), the two alleles at an M-QTL detected in Experiment 1 had opposite effects in different environments. This suggests that the two alleles at these loci responded differently to the environments. Similarly, five HD E-QTL pairs showed opposite AAE effects in different environments. These represented the extreme situation of QTL × environment interaction arising from differential reactions of the identified QTLs to the unspecified environmental factors. Results from experiment 2 suggested that differential responses of the detected QTLs in the sampled environments could be attributable to different cultural conditions and/or abiotic and biotic stresses. For instance, all three types of $QTL \times$ environment interaction detected in experiment 1 were also observed in experiment 2, where they could largely be attributed to a specific environmental factors: soil water status and plant density. Opposite effects of QTLs on PH or HD in aerobic or high density conditions compared to lowland fields have important implications for the development of rice cultivars that can be grown using less irrigation water. Sripongpangkul et al. (2000) reported QTL × environment interaction affecting plant height under different levels of submergence where different sets of M-QTLs and E-QTLs affecting internode and leaf elongation were detected, and the expression of most M-QTLs was much stronger under the more stressful condition.

Importance of epistasis

The importance of epistasis in determining quantitative trait variation has been well demonstrated in several recent QTL mapping studies (Doebley et al. 1995; Lark et al. 1995; Li et al. 1997; Yu et el. 1997; Li et al. 2001; Luo et al. 2001). The large number of E-QTLs identified and the involvement of many M-QTLs in epistasis lend strong support to this notion. Our finding that E-QTLs tended to show a greater level of $QTL \times$ environment interaction than the M-QTLs was expected because E-QTLs should more likely be influenced by environments than M-QTLs. Furthermore, we noted that six HD M-QTLs (QHd1b, QHd1c, QHd2a, QHd6b, QHd8a and QHd9) would have gone undetected in 19 single-environment cases had their involvement in epistasis been unexamined (Table 6). Strong evidence for the presence of epistatic interactions between and among different rice HD M-QTLs and their differential responses to daylength has been clearly demonstrated using near-isogenic lines (Lin et al. 2000).

Trait- and gene-specificity of QTL \times environment interaction

An important finding of this study was that GEIs were trait- and gene (QTL)-specific. As described above, the greater level of GEIs observed for HD in the DH population was reflected by more environment-specific QTLs, greater frequency and magnitude of AE and AAE effects, and more pronounced epistasis for HD than for PH. The level of $QTL \times$ environment interaction varied considerably among the M-QTLs or E-QTLs. Some M-QTLs and E-QTLs were detectable in all or most of the environments and/or showed less variation in their effects. Very often, when detectable in multiple environments, the QTL (phenotypic) effects tended to be in the same direction and showed relatively small variation in magnitude across the environments. These non-environment-specific QTLs included 11 M-QTLs (9 for PH and 2 for HD) and 9 E-QTLs (7 for PH and 2 for HD). These QTLs should be particularly useful in marker-aided manipulation of plant height and flowering time in rice. For those QTLs that were more environment-specific and/ or involved in epistasis, one should be cautious in applying marker-aided selection.

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