

Variation in a single allele drives divergent yield responses to elevated CO₂ between rice subspecies

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Rising atmospheric CO₂ generally increases yield of *indica* rice, one of the two main Asian cultivated rice subspecies, more strongly than *japonica* rice, the other main subspecies. The molecular mechanisms driving this difference remain unclear, limiting the potential of future rice yield increases through breeding efforts. Here, we show that between-species variation in the *DNRI* (*DULL NITROGEN RESPONSE1*) allele, a regulator of nitrate-use efficiency in rice plants, explains the divergent response to elevated atmospheric CO₂ (eCO₂) conditions. eCO₂ increased rice yield by 22.8–32.3% in plants carrying or mimicking the *indica* *DNRI* allele, but only by 3.6–11.1% in plants carrying the *japonica* *DNRI* allele. Rice plants carrying or mimicking the *indica* *DNRI* allele exhibit decreased eCO₂-responsive transcription and protein abundance of *DNRI*, which activates genes involved in nitrate transport and assimilation, driving the increase in plant growth. Our findings identify the *indica* *DNRI* gene as a key breeding resource for sustainably enhancing nitrate uptake and rice yields in *japonica* varieties, potentially contributing to global food security as atmospheric CO₂ levels continue to increase.

Atmospheric CO₂ concentrations have increased from 315 ppm in 1958 to 423 ppm in 2024 and are expected to further increase to 800 ppm by the end of the 21st century due to human activities such as fossil fuel consumption and deforestation^{1,2}. Elevated CO₂ concentration (eCO₂) often stimulates photosynthesis and yields of C3 plants^{3–5}, also known as the CO₂ fertilization effect. Global meta-analyses of free air CO₂ enrichment (FACE) experiments showed that eCO₂ increased photosynthesis by 22% and yield by 14% in rice^{4,6}. Rice sustains approximately

half of the global population and provides about one-fifth of the world's dietary energy supply⁷. Furthermore, global rice yields are predicted to reduce due to global climate warming^{8,9}. In this context, the role of atmospheric changes, particularly the increase in CO₂ levels, becomes a pivotal factor in understanding future trends in rice production and global food security.

Around 90% of rice cultivation occurs in Asia, where two major subspecies, *indica* (Xian) and *japonica* (Geng), each with distinct

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developmental and physiological traits, are cultivated¹⁰. *Japonica* rice, domesticated in the Yangtze River basin around 9000 to 6000 years ago, is predominantly grown and consumed in East Asia, comprising about 40% of the rice cultivation area in China, Japan, and Korea¹¹. In contrast, *indica* rice, first domesticated in the Ganges region between 8500 and 4500 years ago, is more prevalent in most other regions. However, CO₂ fertilization effect varies between *indica* and *japonica*^{6,12–14}, with *indica* varieties showing a more pronounced increase in yield (+20.4%) under eCO₂ conditions compared to *japonica* varieties (+12.7%)¹². The disparity in the CO₂ fertilization between subspecies may strongly affect future global total rice yields; the difference between all rice globally responding to eCO₂ like *japonica* or like *indica* amounts to -109 Tg per year, equal to roughly half of China's annual rice production. However, the underlying mechanisms driving this divergence in response to eCO₂ between rice subspecies, especially at the molecular level, remain largely unexplored.

Nitrogen (N) is a critical nutrient for plant growth, and its availability and uptake significantly affect the response of C3 plants, including rice, to eCO₂ conditions^{3,15,16}. In rice, nitrate (NO₃⁻) and ammonium (NH₄⁺) serve as the primary inorganic N sources¹⁷. Due to nitrification in the rhizosphere, up to 40% of the total N absorbed and utilized by rice is NO₃⁻¹⁸. Notably, *indica* rice varieties generally exhibit higher capacities of both NO₃⁻ uptake and assimilation compared to *japonica*, whereas the two subspecies show similar NH₄⁺ uptake rates^{19,20}. Consistent with these results, we found that *japonica* varieties exhibited significantly lower NO₃⁻ uptake rates that were less sensitive to changes in external CO₂ status compared with *indica* varieties (Supplementary Fig. 1). Furthermore, we observed an interactive effect on NO₃⁻ uptake ($P = 0.001$), but not on NH₄⁺ absorption ($P = 0.373$), between eCO₂ and rice species. The NO₃⁻ uptake rates in *japonica* varieties were less responsive to eCO₂ (+40%) compared to *indica* varieties (+69%) (Supplementary Fig. 1). Based on these observations, we hypothesize that variations in NO₃⁻-use efficiency might contribute to the divergent yield responses to eCO₂ between *indica* and *japonica*.

Recent studies have identified five genes—*OsNRT1.1B*, *OsNR2*, *DULL NITROGEN RESPONSE1 (DNRI)*, *REGULATOR OF N-RESPONSIVE RSA ON CHROMOSOME 10 (RNRI0)*, and *MYB61*—that are involved in NO₃⁻ use efficiency and that exhibit allelic variations between *indica* and *japonica* rice subspecies^{19,21–24} (Supplementary Tables 1, 2). To evaluate the response of genetic variations to eCO₂, we performed an RNA-seq analysis, using the flag leaves of a typical *indica* variety Yangdao 6 (YD6) and a typical *japonica* variety Zhonghua 11 (ZH11) at the heading stage in a FACE experiment (See supplementary method). We found that while eCO₂ (~ +150 ppm) decreased *DNRI* abundance in the leaves of YD6, it did not significantly affect its expression in ZH11 (Supplementary Fig. 2a). In contrast, the responses of the other four genes to eCO₂ were similar between the two varieties (Supplementary Fig. 2b–e). Furthermore, we found that the *DNRI* allele varies between *japonica* and *indica* varieties frequently used in FACE experiments (Supplementary Table 3). Notably, rice varieties with the *indica DNRI* allele were more responsive to eCO₂ (Supplementary Fig. 3).

DNRI, a negative regulator of auxin biosynthesis, has diverged in sequence between the two rice subspecies²³. Plants carrying the *indica DNRI* variant confers reduced *DNRI* mRNA and protein abundance and subsequently increased auxin accumulation, thereby inducing transcriptional activation of genes coding NO₃⁻ uptake and downstream NO₃⁻ assimilation enzymes (Supplementary Fig. 4, Supplementary Table 2), leading to high N-use efficiency (NUE) and grain yield²³. As expected, eCO₂ increased the rice biomass, N uptake and yields more strongly in YD6 than those in ZH11 (Supplementary Fig. 5). These findings suggest that *DNRI* variations could play a significant role in the divergent rice responses to eCO₂ between rice subspecies.

Results

DNRI variation drives divergent yield responses to elevated CO₂

To evaluate whether the *DNRI* variation contributes to the divergent responses to eCO₂ between rice subspecies, we carried out a field experiment within a FACE system to compare typical *japonica* variety ZH11 and its *dnr1* mutants, which mimics the *indica DNRI* allele against a ZH11 background by reducing *DNRI* abundance²³. eCO₂ reduced both *DNRI* transcript and protein abundances in ZH11, and *DNRI* protein levels were undetectable in the *dnr1* mutants under both eCO₂ and aCO₂ conditions, due to the complete loss-of-function nature of the *DNRI* variant (Fig. 1a, b). We found that eCO₂ stimulated the growth of the *dnr1* mutants more strongly than the ZH11 variety (Fig. 1c). At the heading stage, eCO₂ increased the light-saturated net photosynthesis rate more strongly in *dnr1* mutants than in ZH11 (Supplementary Fig. 6a). eCO₂ significantly reduced stomatal conductance (g_s), maximum rate of RuBP carboxylation (V_{cmax}), maximum rate of electron transport driving RuBP regeneration (J_{max}), and non-photochemical quenching (NPQ) (Supplementary Figs. 6, 7). While *dnr1* mutants exhibited higher V_{cmax} and NPQ compared to ZH11, eCO₂ had no effect on the quantum yield of Photosystem II (Y(II)), the photochemical quenching coefficient (qL), or the maximum photochemical quantum yield of PSII (Fv/Fm) (Supplementary Fig. 7). eCO₂ increased leaf area by 40.8% in *dnr1* mutants, compared to only 6.6% increases in ZH11 (Supplementary Fig. 8a). In agreement with previous studies⁶, eCO₂ tended to reduce the N content of flag leaf due to the dilution effect (Supplementary Fig. 8b). eCO₂ increased N uptake more substantially in *dnr1* mutants than in ZH11 (Supplementary Fig. 8c).

At maturity stage, the aboveground biomass, N uptake, and rice yield were all significantly higher in the *dnr1* mutants compared to ZH11 (Fig. 1c–e). The impacts of eCO₂ on these parameters were clearly dependent on the *DNRI* allele. In *dnr1* mutants, eCO₂ stimulated the aboveground biomass, N uptake, and rice yield by 17.8%, 17.8%, and 22.8%, respectively, while it had no significant effects on ZH11 (Fig. 1c–e). Also, eCO₂ stimulated N-use efficiency (NUE) more strongly in *dnr1* than in ZH11 (Fig. 1f).

To test whether *DNRI* variation also causes differential responses to eCO₂ within an *indica* variety, we compared the effects of eCO₂ on the typical *indica* variety, Hua-Jing-Xian 74 (HJX74), with a near-isogenic line (NIL) carrying the *japonica DNRI* allele for two years. The *indica* HJX74 exhibited lower *DNRI* transcripts and protein abundance compared to NIL, and the changes induced by eCO₂ were stronger in HJX74 (Supplementary Fig. 9a, b). More importantly, eCO₂ stimulated rice plant growth parameters (i.e., photosynthesis rate, leaf area, biomass, N uptake, and NUE) more strongly in HJX74 than in NIL (Supplementary Figs. 9, 10). At maturity, eCO₂ significantly increased rice yield by 27.6–32.3% in HJX74, but only by 3.6–11.1% in NIL (Supplementary Figs. 9j, 10h). Taken together, our findings indicate that the variation in the *DNRI* allele plays a crucial role in driving the divergent responses to eCO₂ between rice subspecies.

DNRI variation affects response of N metabolism to eCO₂

To determine how *DNRI* variation affects the response of N metabolism to eCO₂, we conducted a 14-day hydroponic experiment in walk-in growth chambers, administering either aCO₂ or eCO₂ treatments to ZH11 and *dnr1* mutants, as well as HJX74 and NIL. Similar with results of FACE experiment, eCO₂ enhanced aboveground biomass, especially in *dnr1* mutants (Fig. 2a, c). eCO₂ also enlarged root system architecture (RSA), including total root length, total area, and the total number of root tips (Fig. 2b, d–f). While eCO₂ increased ¹⁵NO₃⁻ uptake rates in both ZH11 and *dnr1* mutants (Fig. 2g), the positive effect of eCO₂ was more pronounced in the *dnr1* mutants (Fig. 2). Similarly, eCO₂ promotes aboveground biomass, RSA and ¹⁵NO₃⁻ absorption rate more strongly in HJX74 than in NIL (Supplementary Fig. 11a–g). These effects are attributed to increased IAA content under eCO₂ treatment, caused

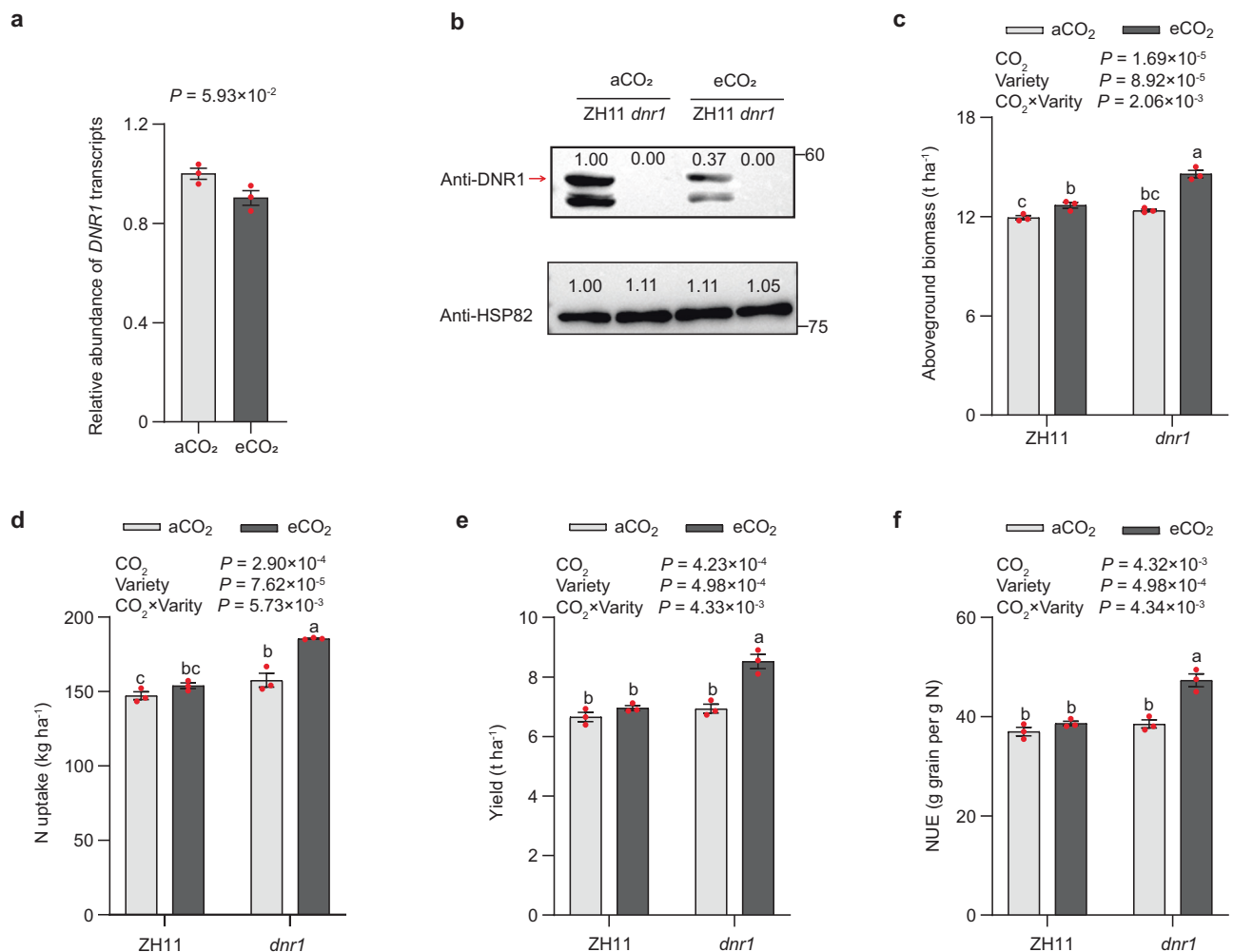


Fig. 1 | Different responses of *japonica* variety Zhonghua 11 (ZH11) and its *dnr1* mutants mimicking the *indica* DNRI allele to elevated CO₂. a DNRI transcript abundance in ZH11 shoots under ambient CO₂ (aCO₂) and elevated CO₂ (eCO₂) conditions. Transcript abundance was measured relative to ZH11 under aCO₂ (set to 1). Data are mean ± s.e.m. ($n = 3$ biological replicates). P -value was generated from two-sided Student's t tests. **b** DNRI protein abundance in the shoots of ZH11 and *dnr1* under aCO₂ and eCO₂ conditions. HSP82 serves as a loading control. The red

arrow indicates the DNRI bands. Data are representative of three independent experiments, with similar results. Aboveground biomass (**c**), N uptake (**d**), yield (**e**) and N-use efficiency (**f**) under aCO₂ and eCO₂ conditions were measured at maturity. ZH11 and *dnr1* indicate *japonica* variety Zhonghua 11 and its *dnr1* mutants, respectively. **c–f** Data are mean ± s.e.m. ($n = 3$ biological replicates). P -values were generated from two-way ANOVA. Different letters indicate significant differences among treatments ($P < 0.05$). Source data are provided as a Source Data file.

by the inhibition of DNRI expression, particularly in the *dnr1* mutant and HJX74 (Fig. 2h; Supplementary Fig. 11h).

Additionally, we conducted hydroponic experiments using either NO₃⁻ (KNO₃) or NH₄⁺ ((NH₄)₂SO₄) as the N source for ZH11 and *dnr1* plants. *dnr1* plants consistently showed a higher responsiveness to variable CO₂ levels in both NO₃⁻ and NH₄⁺ conditions compared to ZH11 (Supplementary Fig. 12). However, the responses of aboveground and root biomass in both ZH11 and *dnr1* were more pronounced under eCO₂ when NO₃⁻ was provided, compared to NH₄⁺ (Supplementary Fig. 12). Therefore, it is reasonable to conclude that DNRI plays a crucial role in mediating plant growth in response to varying CO₂ levels, particularly with NO₃⁻.

DNRI acts as an antagonist to auxin biosynthesis, triggering AUXIN RESPONSE FACTORS (OsARF6 and OsARF17)-mediated activation of genes associated with NO₃⁻ uptake and metabolism, thereby contributing to enhanced NUE and grain yield²³ (Supplementary Fig. 1). To determine whether eCO₂ influences NO₃⁻ metabolism by modulating DNRI-mediated auxin homeostasis, we conducted RT-qPCR analysis of key genes involved in the DNRI-auxin-N pathway, including *OsARF6*, *OsARF17*, *OsNRT1.1B*, *OsNRT2.3a*, *OsNPF2.4*, and *OsNIA2*. As anticipated, eCO₂ significantly increased transcription levels of

OsARF6, *OsARF17*, *OsNRT1.1B*, *OsNRT2.3a*, *OsNPF2.4*, and *OsNIA2* (Figs. 2h and 3a–f). Correspondingly, nitrate reductase (NR) activity was markedly enhanced under eCO₂ conditions (Supplementary Fig. 13a). Notably, eCO₂ stimulated the expression of these genes in the *dnr1* mutants more strongly than in ZH11, resulting in higher NO₃⁻ uptake rate and NR activity (Fig. 3; Supplementary Fig. 13a). Similar trends were observed in HJX74 and NIL, with or without eCO₂ treatment (Supplementary Figs. 13b and 14).

Next, we investigated whether eCO₂ could stimulate NO₃⁻ uptake and assimilation through the involvement of OsARFs, which act as downstream transcription factors of auxin homeostasis mediated by DNRI. Firstly, we conducted a time course assessment of the expression levels of *OsARF6*, *OsARF17*, *OsNRT1.1B*, *OsNRT2.3a*, *OsNPF2.4*, and *OsNIA2* and found that the expression patterns of these four N-related genes and *OsARFs* are aligned. Specifically, when *OsARFs* expression is strongly induced by eCO₂ (30 min to 2 h), the expression of *OsNRT1.1B*, *OsNRT2.3a*, *OsNPF2.4*, and *OsNIA2* also significantly increases, suggesting that *OsARFs* regulates these genes in response to eCO₂ (Supplementary Fig. 15). Interestingly, under eCO₂ conditions, *OsNRT1.1B* shows a quicker response, which may be due to its direct response to CO₂ concentration changes.

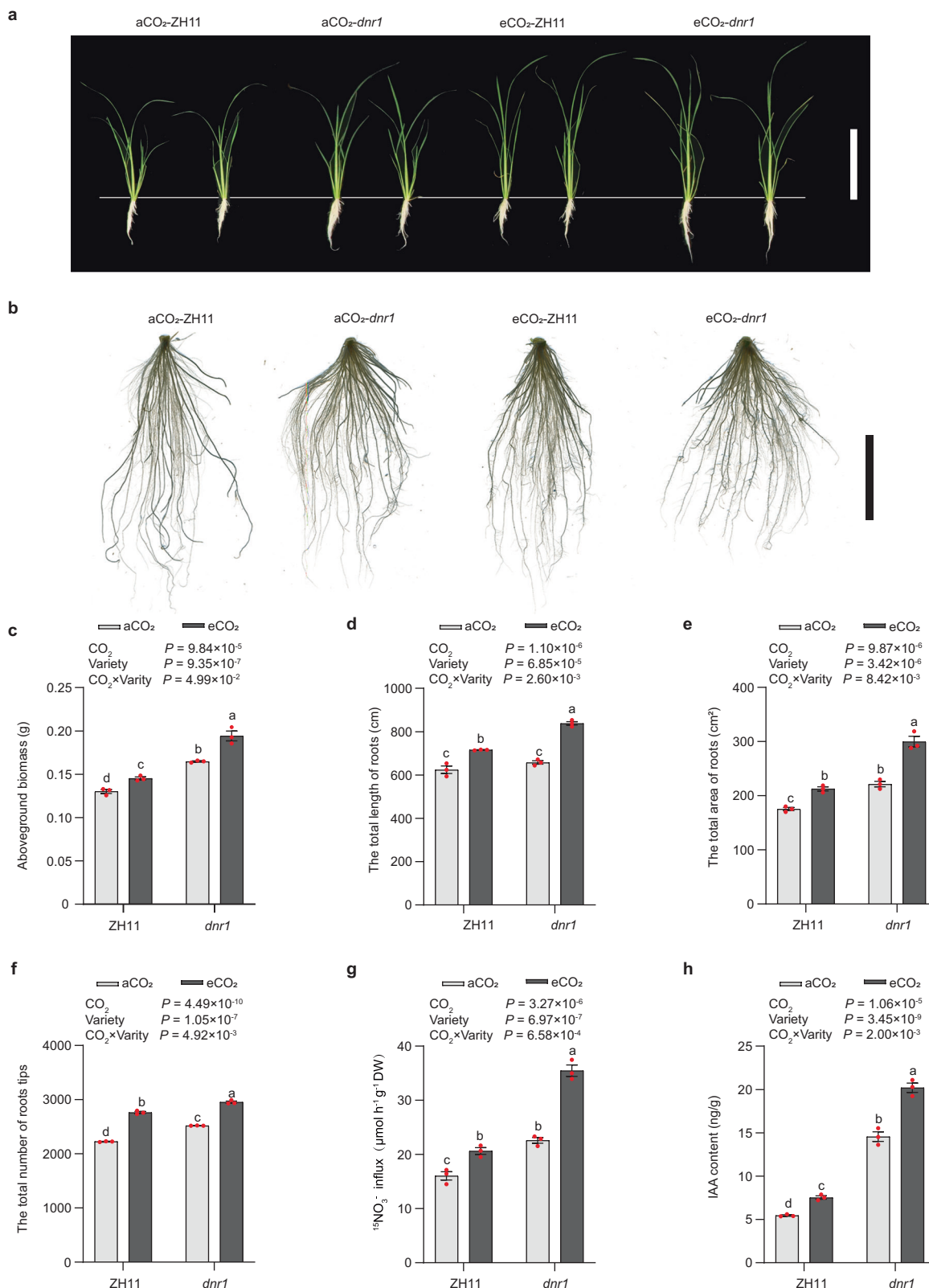


Fig. 2 | Differences in plant growth and root developmental plasticity in response to elevated CO₂ between japonica variety Zhonghua 11 (ZH11) and its *dnr1* mutants mimicking the *indica* DNRI allele. **a, **b** 14-day-old japonica variety Zhonghua 11 (ZH11) and its *dnr1* mutants (*dnr1*) rice plants grown under ambient CO₂ (aCO₂) and elevated CO₂ (eCO₂) conditions, respectively. Morphology of plants (**a**) and root systems (**b**). **a** Scale bar, 20 cm. **b** Scale bar, 5 cm. **c** Aboveground**

biomass. Root statistics of total length of visible roots (**d**), total area of visible roots (**e**), and number of root tips (**f**). ¹⁵N₃ uptake rates (**g**) and root free IAA content (**h**) of japonica variety ZH11 and *dnr1* under aCO₂ and eCO₂ conditions. **c–h** Data are mean ± s.e.m. (*n* = 3 biological replicates). *P*-values were generated from two-way ANOVA. Different letters indicate significant differences among treatments (*P* < 0.05). Source data are provided as a Source Data file.

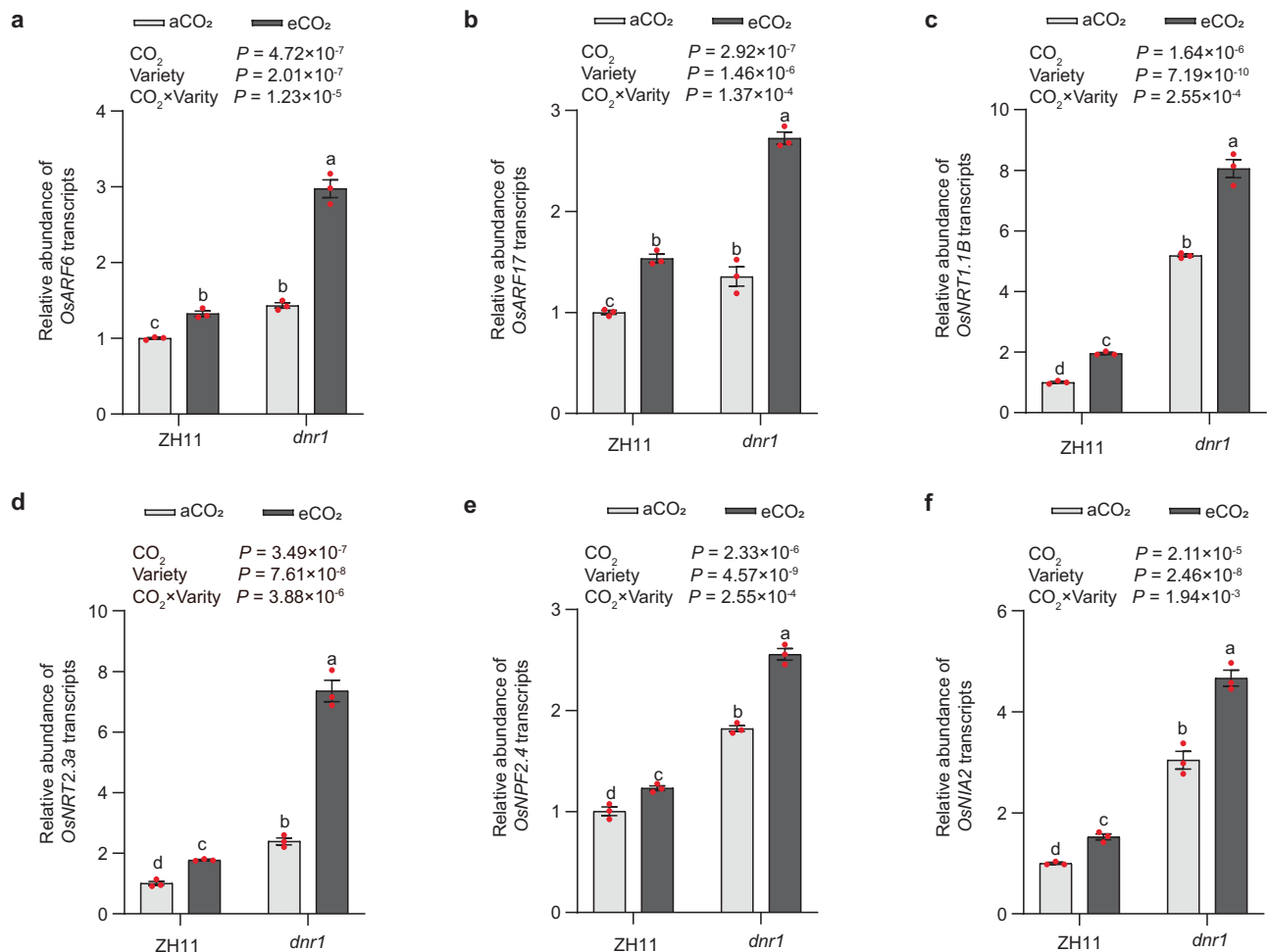


Fig. 3 | Elevated CO₂ enhances NO₃⁻ metabolism via DNRI-mediated auxin homeostasis in *japonica* variety Zhonghua 11 (ZH11) and its *dnr1* mutants mimicking the *indica* DNRI allele. Root mRNA abundances of *OsARF6* (a) and *OsARF17* (b) grown under ambient CO₂ (aCO₂) and elevated CO₂ (eCO₂) conditions, relative to ZH11 under aCO₂ (set to 1). Root mRNA abundances of *OsNRT1.1B* (c) and *OsNRT2.3a* (d) relative to ZH11 under aCO₂ (set to 1). Shoot mRNA abundances of

OsNPF2.4 (e) and *OsNIA2* (f) relative to ZH11 under aCO₂ (set to 1). ZH11 and *dnr1* indicate *japonica* variety Zhonghua 11 and its *dnr1* mutants, respectively. a–f Data are mean ± s.e.m. ($n = 3$ biological replicates). P -values were generated from two-way ANOVA. Different letters indicate significant differences among treatments ($P < 0.05$). Source data are provided as a Source Data file.

To test this hypothesis, we used HJX74 and a single-segment substitution line (SSSL-064), generated by crossing IRAT261 (donor parent) with HJX74 (recurrent parent), which incorporates a chromosome segment containing *OsNRT1.1B* from *japonica* IRAT261 into the HJX74 genetic background to investigate the response of *OsNRT1.1B* to eCO₂. Both HJX74 and SSSL-064 exhibited increased biomass and ¹⁵NO₃⁻ uptake under eCO₂ compared to aCO₂, with HJX74 showing a slightly more pronounced increase (Supplementary Fig. 16). These results indicate that *OsNRT1.1B* itself can influence ¹⁵NO₃⁻ absorption and thereby affect growth to some extent in response to eCO₂.

Secondly, to further illustrate the regulatory effects of OsARF6 and OsARF17 under aCO₂ and eCO₂ conditions, we performed ChIP-qPCR assays. These assays indicated that eCO₂ enhanced the enrichment of TGTCTC/GAGACA motif-containing fragments from the promoters of *OsNRT1.1B*, *OsNRT2.3a*, *OsNPF2.4*, and *OsNIA2* compared to the aCO₂ treatment (Fig. 4a–d; Supplementary Fig. 17a–f). The following in vitro transient transactivation assays revealed that eCO₂ increased the transcriptional activation capacities of OsARF6 and OsARF17 towards their downstream genes, *OsNRT1.1B*, *OsNRT2.3a*, *OsNPF2.4*, and *OsNIA2*, when compared to aCO₂ treatment (Fig. 4e, f; Supplementary Fig. 17g, h). Accordingly, under both aCO₂ and eCO₂ conditions, the expression levels of *OsNRT1.1B*, *OsNRT2.3a*, *OsNPF2.4*, and *OsNIA2* were upregulated in the *OsARF6* or *OsARF17*

overexpression lines within the ZH11/*pACT::DNRI-Flag* background (Fig. 4g–j). These results further confirm the transactivation activities of OsARFs towards N-related genes in rice protoplasts (Fig. 4e, f; Supplementary Fig. 17g, h).

Finally, we tested whether the altered transcriptional activation capacities of OsARF6 and OsARF17 led to changes in NO₃⁻ uptake in response to eCO₂. Under eCO₂, the over-expression of *OsARF6* or *OsARF17* in the ZH11/*pACT::DNRI-Flag* background significantly restored the insensitivities of NO₃⁻ absorption caused by auxin depletion (Fig. 4k). This suggests that eCO₂ promotes NO₃⁻ uptake and assimilation by enhancing the transcriptional activation capacities of OsARFs.

DNRI variation affects response of photosynthetic genes to eCO₂

As a crucial substrate for photosynthesis, CO₂ actively participates in the photosynthetic processes of plants. Thus, we compared the transcription and protein levels of core elements in photosystems (*OsPsaB*, *OsPsaD*), and rate-limiting factors in the Calvin-Benson cycle (*OsRbcS*, *OsRbcL* and *OsSBPase*) (Supplementary Table 4), in both ZH11 and *dnr1* mutants with or without CO₂ treatment. Both transcription and protein levels of *OsPsaB*, *OsRbcS* and *OsSBPase* were significantly higher in *dnr1* mutants compared to ZH11, and the changes induced by eCO₂ were

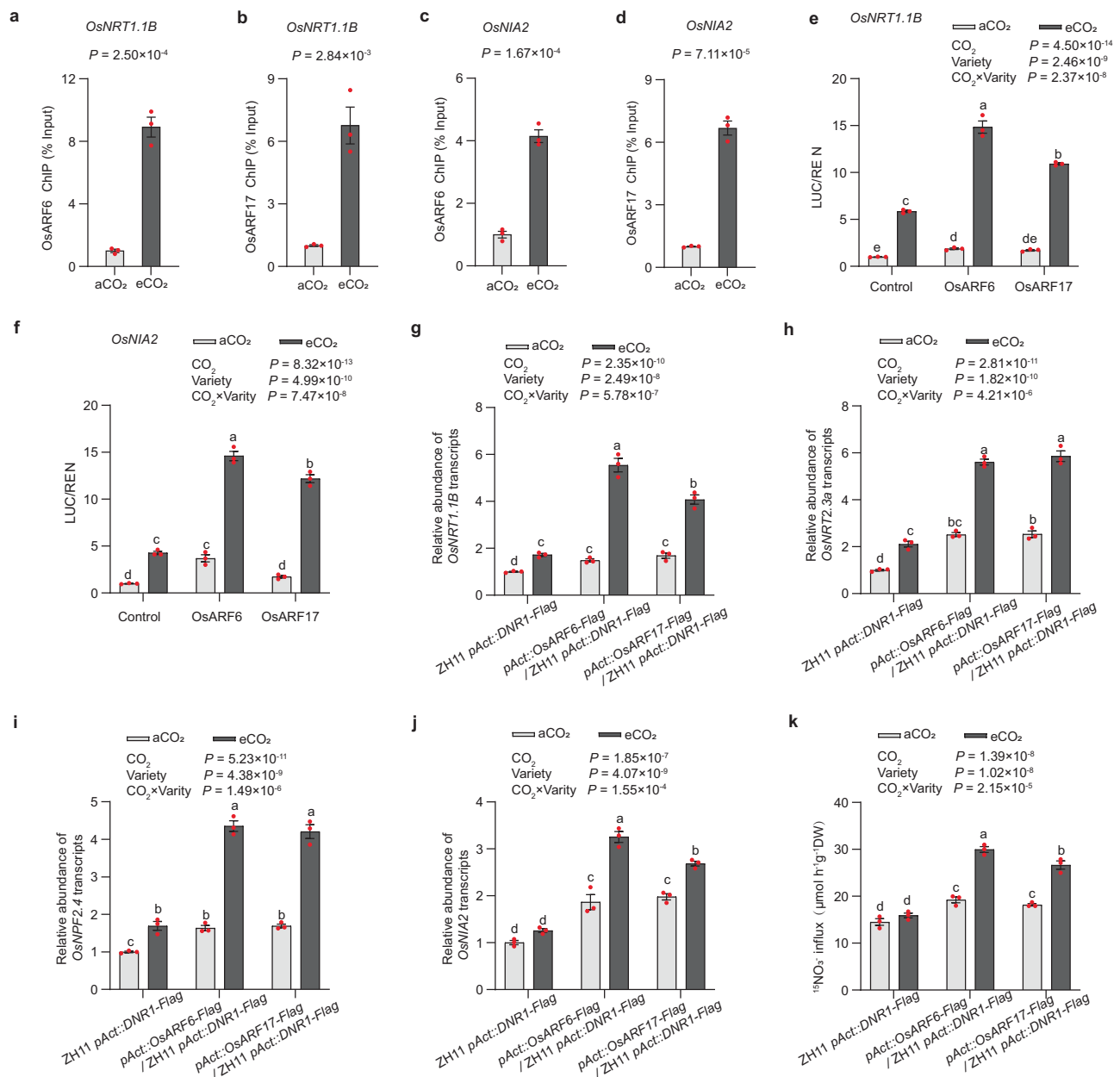


Fig. 4 | The effect of elevated CO₂ on the expression of N metabolism-related genes and NO₃⁻ uptake is mediated by DNRI-OsARFs module. Extent of OsARF6 and OsARF17-mediated ChIP-qPCR enrichment (relative to Input) of TGTCTC-containing promoter fragments from *OsNRT1.1B* (a, b) and *OsNIA2* (c, d) under ambient CO₂ (aCO₂) and elevated CO₂ (eCO₂) conditions. Data are mean ± s.e.m. ($n = 3$ biological replicates). P -values were generated from two-sided Student's t tests. OsARF6 and OsARF17 activate *OsNRT1.1B* (e) and *OsNIA2* (f) promoter-LUC fusion constructs in transient transactivation assays. The LUC/REN activity obtained from a co-transfection with an empty effector construct and indicated

reporter constructs under ambient CO₂ (aCO₂) was set to 1. Root mRNA abundances of *OsNRT1.1B* (g) and *OsNRT2.3a* (h) relative to ZH11 *pAct::DNRI-Flag* under aCO₂ (set to 1). i, j Shoot mRNA abundances of *OsNPF2.4* (i) and *OsNIA2* (j) relative to ZH11 *pAct::DNRI-Flag* under aCO₂ (set to 1). k Root ¹⁵NO₃⁻ uptake rate of *OsARF6* and *OsARF17* overexpression lines in the ZH11 *pAct::DNRI-Flag* background grown under aCO₂ and elevated CO₂ (eCO₂) conditions, respectively. e–k Data are mean ± s.e.m. ($n = 3$ biological replicates). P -values were generated from two-way ANOVA. Different letters indicate significant differences among treatments ($P < 0.05$). Source data are provided as a Source Data file.

more pronounced in *dnr1* mutants (Fig. 5). However, although the expression levels of *OsPsbA*, *OsRbcS4* and *OsRbcL* slightly increased in *dnr1*, they appear to be unaffected by eCO₂ (Supplementary Fig. 18). Similar results were found for HJX74 and NIL (Supplementary Fig. 19), confirming that the abundance patterns of photosynthetic genes induced by eCO₂ in ZH11 and *dnr1* mutants were shared by these varieties.

We found that OsARF6 and OsARF17 do not possess transcriptional activation abilities for these genes under both aCO₂ and eCO₂ conditions (Supplementary Fig. 20a), suggesting that eCO₂ promotes

the accumulation of photosynthesis-related proteins and enhances photosynthesis independent on OsARFs. Additionally, examining the expression levels of these genes in plants overexpressing *OsARF6* and *OsARF17* within the ZH11/*pAct::DNRI-Flag* background revealed no changes compared to ZH11/*pAct::DNRI-Flag* with either aCO₂ or eCO₂ treatment (Supplementary Fig. 20b–g). This suggests that DNRI does not regulate photosynthetic efficiency through the transcriptional activation of OsARF6 and OsARF17.

We conducted RNA sequencing analysis on ZH11 and the *dnr1* mutant under both aCO₂ and eCO₂ conditions and identified 397 target

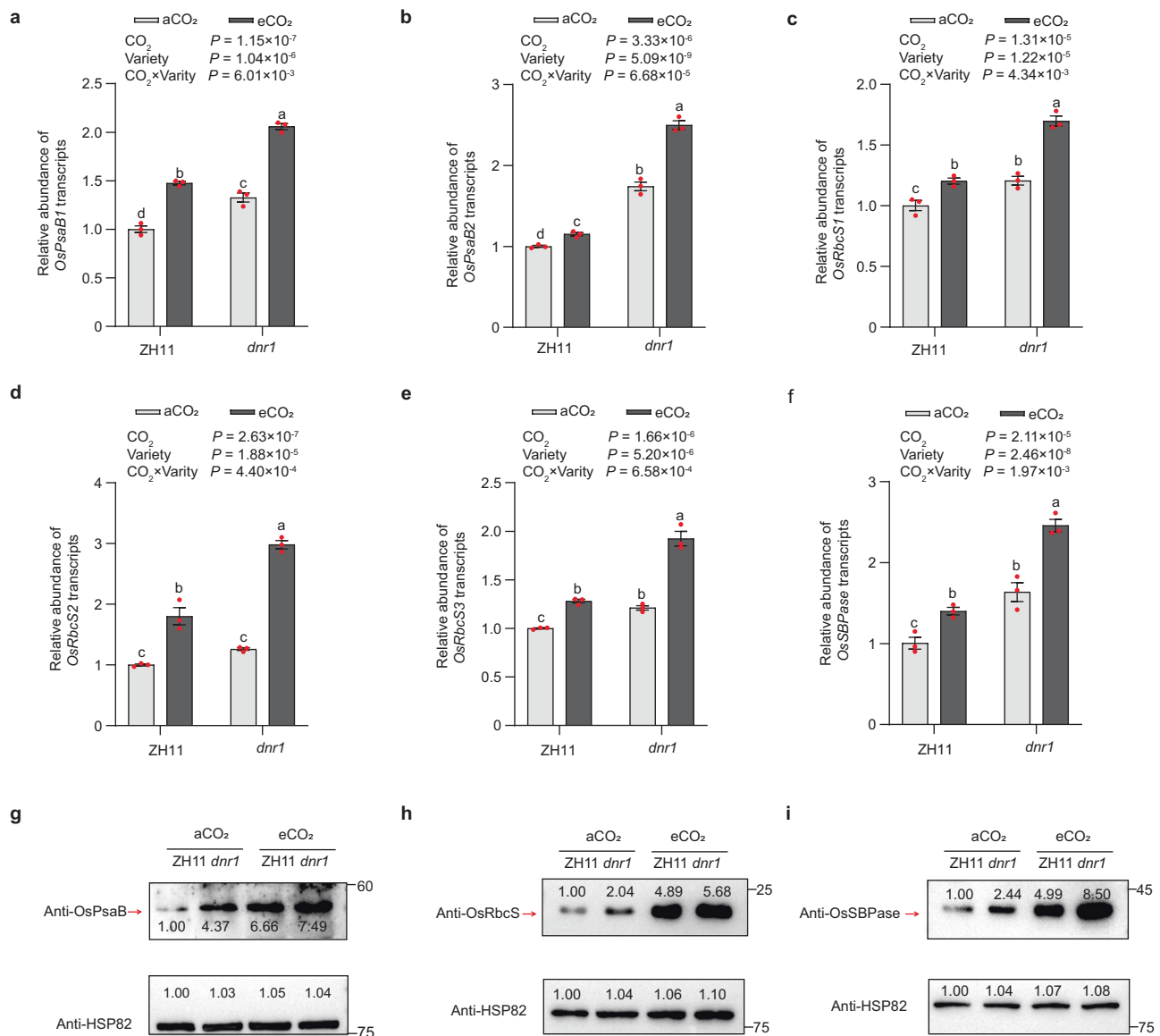


Fig. 5 | The impact of elevated CO_2 on DNRI-regulated abundances of photosynthetic genes in *japonica* variety Zhonghua 11 (ZH11) and its *dnr1* mutants mimicking the *indica* DNRI allele. Shoot mRNA abundances of *OsPsaB1* (a), *OsPsaB2* (b), *OsRbcS1* (c), *OsRbcS2* (d), *OsRbcS3* (e), and *OsSBPase* (f) grown under ambient CO_2 (a CO_2) and elevated CO_2 (e CO_2) conditions, respectively, relative to Zhonghua 11 under a CO_2 (set to 1). Data are mean \pm s.e.m. ($n = 3$ biological replicates). P -values

were generated from two-way ANOVA. Different letters indicate significant differences among treatments ($P < 0.05$). *OsPsaB* (g), *OsRbcS* (h) and *OsSBPase* (i) protein abundances in shoots. *HSP82* serves as a loading control. The red arrows indicate the *OsPsaB*, *OsRbcS* and *OsSBPase* bands, respectively. Data are representative of three independent experiments, with similar results. Source data are provided as a Source Data file.

genes regulated by both CO_2 and DNRI. Among these, 9 transcription factors are upregulated by both e CO_2 and null-DNRI allele, and 4 transcription factors are downregulated by both e CO_2 and null-DNRI allele (Supplementary Table 5). These 13 transcription factors may serve as potential candidates for regulating photosynthetic efficiency in response to e CO_2 and DNRI interactions, offering promising avenues for future research. Overall, e CO_2 enhances photosynthetic efficiency by improving the C and N cycles through various mechanisms.

Discussion

We present evidence of the key role of DNRI in driving the divergent responses of *indica* and *japonica* rice varieties to e CO_2 (Fig. 6). e CO_2 likely influences DNRI indirectly through changes in N status. e CO_2 can increase photosynthesis and plant growth, thereby raising the demand for nitrogen. This, in turn, could lead to a decrease in DNRI abundance, stimulating N uptake. Specifically, plants carrying the *japonica* DNRI

allele, which exhibit higher DNRI abundance leading to reduced auxin accumulation, key traits such as biomass, nitrogen content, and yield respond relatively weakly to e CO_2 . Conversely, in *dnr1* mutants, e CO_2 decreases the expression of DNRI and the following transcriptional activation of OSARF6 and OSARF17, which in turn upregulates genes associated with NO_3^- uptake and assimilation (*OsNRT1.1B*, *OsNRT2.3a*, *OsNIA2*, and *OsNPF2.4*). Consequently, N uptake and assimilation increased, resulting in elevated N content in rice plants, which enhanced the photosynthetic capacity of the plants under e CO_2 concentrations. Additionally, previous studies have shown that IAA levels rise under e CO_2 ^{25–27}. Our results suggest that the inhibition of DNRI by e CO_2 may also contribute to this increase in IAA content, which further enhances our understanding of the relationship between e CO_2 and IAA homeostasis.

Previous studies have extensively explored the impacts of e CO_2 on leaf photosynthesis^{28–30}. A meta-analysis of 20-year rice FACE

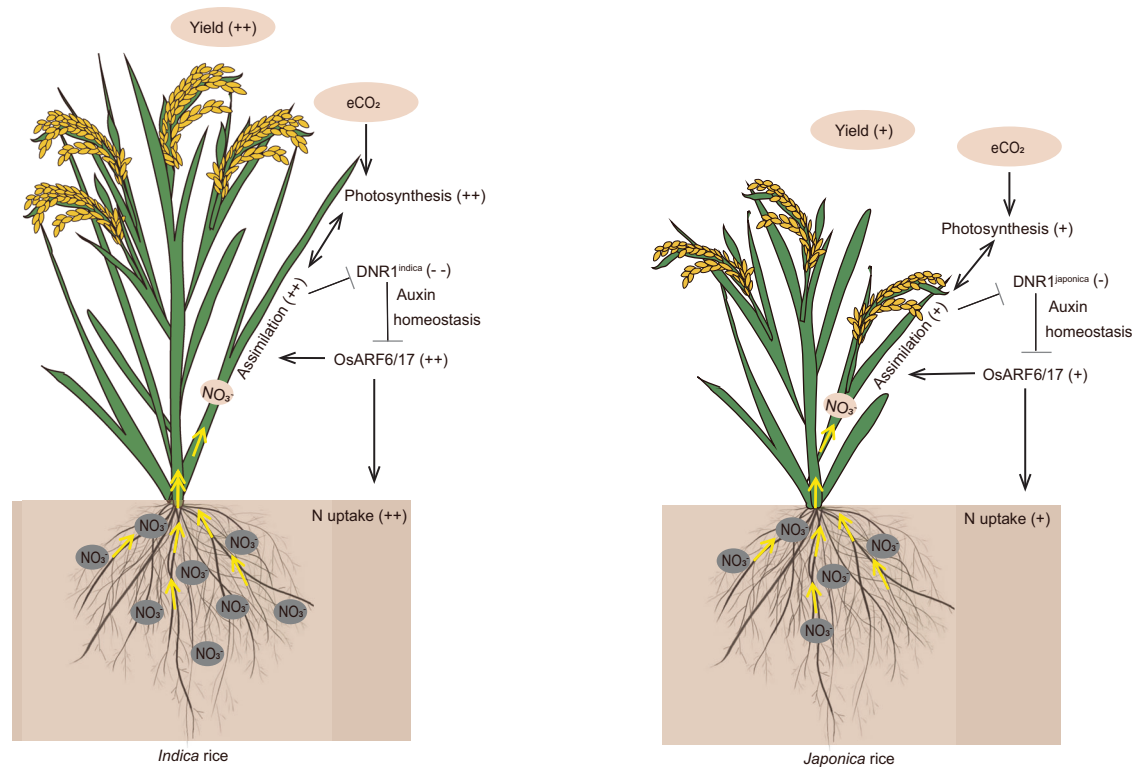


Fig. 6 | Schematic overview of the role of DNR1 in driving the divergent responses of *indica* and *japonica* rice to elevated CO₂. Rising CO₂ concentrations increase photosynthesis and rice yields by affecting a range of genes involved in

uptake, transport, and assimilation of nitrate, which in turn are regulated by DNR1. +, ++, -, and -- indicate positive, strongly positive, negative, and strongly negative, respectively.

studies revealed that eCO₂ significantly increased the light-saturated photosynthetic rate of leaves but reduced g_s , V_{cmax} , and J_{max} ⁶, consistency with our study. However, the response of photosynthetic genes to elevated CO₂ has rarely been investigated. In this study, we showed that eCO₂ promotes the abundance of photosynthetic genes (*OsPsaB*, *OsRbcS* and *OsSBPase*) more strongly in plants carrying the *indica* DNR1 allele, leading to higher yield response to eCO₂. Interestingly, DNR1 mediated the regulation of photosynthetic genes expression independently of OsARFs. Further research is needed to elucidate why DNR1 influences C and N cycles through different mechanisms.

Together, our findings demonstrate that DNR1 plays a crucial role in coordinating N metabolism and C fixation to enhance plant growth in response to eCO₂. This insight provides a valuable breeding strategy for adapting to increasing atmospheric CO₂ levels by modulating DNR1, which acts as a pivotal bridge in this process. Our findings also provide insights into the relationship between N metabolism and the rate of photosynthetic C fixation in plants, particularly in the context of molecular coupling mechanisms. This study identifies molecular mechanisms driving CO₂ effects among rice subspecies.

Our findings corroborate previous studies showing that in seed crops, sink capacity is a critical factor limiting yield response to eCO₂^{4,31,32}. Specifically, we found that eCO₂ increased sink capacity more strongly in HJX74 than in NIL, and more strongly in *dnr1* mutants than in ZH11, resulting in higher yields (Supplementary Table 6). Similarly, eCO₂ increases the spikelet numbers per panicle more strongly in *indica* rice as compared to *japonica* rice, resulting in a heightened overall yield response^{12,14}. Importantly, this sink capacity is intricately regulated by N availability. Indeed, by increasing NO₃⁻ use efficiency, *dnr1* stimulates sink capacity, as evidenced by increased panicle size, spikelet number, and seed size²³. Furthermore, a previous FACE study also showed that effects of eCO₂ on rice yield are affected by N application rate^{33,34}. Collectively, these results indicate that the

absorption and utilization of N by rice plants is a pivotal factor influencing yield responses to eCO₂.

Two limitations of our experiment should be noted. First, we only used ZH11 as the typical *japonica* and YD6 and HJX74 as the typical *indica*. While more recipient varieties would strengthen the robustness of our results, the layout of the FACE experiment did not allow for testing more varieties. However, YD6 has been often utilized in other FACE experiments¹⁴. In our experiments, both YD6 and HJX74 exhibited similar responses to eCO₂ levels. Additionally, ZH11's response aligns with the average response observed in other FACE experiments involving *japonica* rice varieties¹⁴. These results strongly suggest that the chosen cultivars are broadly representative for *japonica* and *indica*. Second, although data from the FACE experiment have been used as a basis for hydroponic cultivation, the different environmental conditions between the two experiments may have affected rice growth and eCO₂ responses. Unfortunately, this issue is difficult to avoid. On the other hand, the results from our FACE experiment and hydroponic culture experiment consistently support the findings regarding nitrogen absorption and plant growth, suggesting that our findings are robust.

Optimizing yield response to rising atmospheric CO₂ is crucial in mitigating the anticipated supply-demand shortfall this century due to rice yield loss of climatic warming and demand increase^{4,35}. Field experiments have revealed large variation in crop productivity and quality responses to eCO₂ between rice varieties, underscoring the importance of genetic variation in breeding for enhanced productivity and quality under eCO₂ conditions^{4,36,37}. Our study builds on this understanding by showing that DNR1-mediated variations in NO₃⁻ utilization largely explains the difference in yield responses to eCO₂ between rice subspecies.

Importantly, our phylogenetic analysis of ~3000 rice accessions showed that *indica* and *japonica* DNR1 alleles belong to two separate clades^{23,24}. Haplotype analysis of the DNR1 gene of these varieties

revealed four distinct haplotypes (Hap. I-IV). Notably, 98.1% of the *indica* subpopulation belongs to Hap. I, while 75.7% and 22.2% of the *japonica* subpopulation belongs to Hap. II and Hap. III, respectively²⁴, demonstrating consistent differentiation across existing varieties. This divergence may be attributed to high-fertilizer breeding conditions that have led to the effective utilization of the *indica*-type *DNRI*, while it remains underutilized in *japonica* rice. Together, these results suggest that the CO₂ fertilization effect for the vast majority *japonica* varieties can be increased by manipulating *DNRI*.

As *japonica* rice accounts for ~15% of global rice production³⁸, breeding efforts focused on NUE increase are key. Other breeding approaches may also show potential. For instance, eCO₂ increased yield by up to 30.3% in a low-yielding old *japonica* variety, suggesting that other loci are also important in determining the rice response to elevated CO₂. Together these results underscore the importance of rice varieties and agronomic practices with high NUE to improve food security as atmospheric CO₂ concentrations continue to increase.

Methods

Field experiment

FACE system. In 2021, we established a Free Air CO₂ Enrichment (FACE) system in Baolin village (31.9°N, 119.5°E), Yanling Town, Danyang City, Jiangsu Province, China. The FACE system comprises six octagonal rings, each 8 m in diameter. Three rings were designated for ambient atmospheric CO₂ concentration treatment (aCO₂), and three for elevated atmospheric CO₂ treatment (eCO₂). To prevent crossover effects, a minimum distance of 25 m was maintained between any aCO₂ and eCO₂ rings. At the center of each ring, a monitoring system was installed to track CO₂ concentration, air temperature, and wind speed. Within each eCO₂ ring, eight CO₂ sensors (VC2008T, SenseAir, Sweden) were placed above the rice canopy, evenly distributed within a 4-m circle at the ring's center.

In line with IPCC¹ predictions, atmospheric CO₂ concentrations are expected to reach between 500 (SSP5) and 800 ppm (SSP3) by 2100. Hence, we set eCO₂ treatments to mimic end-of-century conditions at ~550 ppm. PVC emission tubes, arranged to form an octagon around each eCO₂ ring at 50 cm above the rice canopy, facilitated CO₂ distribution. Considering operational costs, pure CO₂ gas was released only during the daytime (5:00 a.m. to 7:00 p.m.). This release was regulated automatically based on wind direction and speed and was halted when wind speeds exceeded 5 m s⁻¹. The average daytime CO₂ concentrations were maintained at 560 ± 19 ppm in eCO₂ treatments and 387 ± 10 ppm in aCO₂ treatments. Details of climate condition, soil properties, and FACE system can be found in Qian et al.³⁹.

Experimental design. To determine whether variation in the *DNRI* allele drives the divergent responses to eCO₂ between rice subspecies, we planted a typical *japonica* variety, Zhonghua 11 (ZH11), with its *dnr1* mutants, which mimics the *indica* *DNRI* allele in ZH11 background by reducing *DNRI* abundance in 2023 and 2024. We also planted a typical *indica* variety, Hua-Jing-Xian 74 (HJX74), with a near-isogenic line (NIL) carrying the *japonica* *DNRI* allele in 2022 and 2023. Details of these plant materials can be found in Zhang et al.²³. The plots for each treatment measured 1.5 m × 2 m. Two healthy rice seedlings per hill were transplanted at planting spaced at 15 cm × 25 cm intervals. N fertilizer, in the form of urea, was applied at a total rate of 180 kg N ha⁻¹. The N fertilizers were divided into three applications: 40% at soil tillage, 30% at the tillering stage, and the remaining 30% at the jointing stage. Phosphorus fertilizer, at a rate of 120 kg P₂O₅ ha⁻¹, was applied at soil tillage, while potassium fertilizer, at 160 kg K₂O ha⁻¹, was split equally between the soil tillage and jointing stages. All other agronomic practices were conducted in accordance with local agricultural recommendations.

Sampling and measurements. Light response of net photosynthesis (An) and stomatal conductance (gs) of flag leaves were measured at 11 photosynthetically active radiation (PAR) levels (in decreasing order of 1800, 1500, 1200, 1000, 800, 600, 400, 200, 100, 50 and 0 μmol m⁻² s⁻¹) by a portable photosynthesis system (LI-6800, LI-COR, Lincoln, America) between 8:30 and 11:30 a.m. Air temperature and relative humidity in the chamber were set at 35 °C, and 60%, respectively. CO₂ concentrations were set at 400 ppm for the aCO₂ and 550 ppm for the eCO₂ treatment.

We used the rapid A - Ci response (RACiR) technique to obtain the CO₂ response of net photosynthesis rate (An) of flag leaves by a portable photosynthesis system (LI-6800, LI-COR, Lincoln, America) between 8:30 and 11:30 a.m according to Joseph et al.⁴⁰. We set the initial and final CO₂ concentration at 50 and 1200 ppm respective, and ramp rates at 300 ppm min⁻¹. Photosynthetic active radiation, air temperature and relative humidity in the chamber were set at 1800 μmol m⁻² s⁻¹, 35 °C, and 60%, respectively. Then, the resulting functional relationship (A-Ci curves) was used to estimate the *V*_{max} and *J*_{max} FvCB model.

Chlorophyll fluorescence parameters photosystem II quantum yield (Y(II)), non-photochemical quenching (NPQ), photochemical quenching coefficient (qL), and maximum photochemical quantum yield of Photosystem II (Fv/Fm) of flag leaves were measured using the FluorPen (FluorPen FP110 - LM/D, Photon Systems Instruments, Czech) after 1-h period of darkness.

For each plot, three hills were selected to measure leaf area using a table leaf area instrument (LI-3100C, LI-COR, America). Rice plants were harvested at the heading stage and the mature stage, oven-dried at 70 °C to obtain a constant weight and then crushed. The nitrogen content was measured using an elemental analyser (vario PYRO, Elemental, German), and N uptake was calculated by multiplying the aboveground biomass by the N content. NUE is defined as grain weight divided by nitrogen supply⁴¹.

RNA-seq analysis. Total RNAs were extracted from heading stage rice flag grown in field under the FACE system using the QIAGEN RNeasy plant mini kit (QIAGEN, 74904) following the manufacturer's instructions. Three replicate RNA-seq libraries were prepared from YD6 and ZH11 plants under aCO₂ or eCO₂, respectively. A total of the four libraries were sequenced separately using the Illumina Novaseq platform. Raw sequencing reads were cleaned by removing adaptor sequences, reads containing poly-N sequences, and low-quality reads. Approximately 44,226,120 clean reads were mapped to the Nipponbare reference genome using Hisat2 v2.0.5⁴². After data were mapped, normalization was performed and then FPKM (fragments per kilobase per million mapped reads) was calculated using RESM software⁴³. As previously described⁴⁴, a false discovery rate (FDR) < 0.05 and absolute value of log₂ ratio ≥ 2 were used to identify differentially expressed genes in YD6 and ZH11 samples under aCO₂ or eCO₂, respectively.

Hydroponic experiment

Plant materials. The plant materials including a pair of near isogenic materials of NIL-*DNRI*^{HJX74} (HJX74) and NIL-*DNRI*^{RAP9} (NIL), as well as *pAct::DNRI-Flag* and *dnr1* under Zhonghua 11 (ZH11) background, and *pAct::OsARF6-Flag* and *pAct::OsARF17-Flag* under ZH11/*pAct::DNRI-Flag*. Details of these plant materials can be found in Zhang et al.²¹ and Huang et al.²².

Hydroponic culture. Seeds were soaked in 20% sodium hypochlorite solution for 30 min for disinfection and selected with uniform growth for further analyses as described previously⁴⁵. The 7-day-old seedlings were transferred to PVC pots containing 10 L of nutrient solution (1.25 mM NH₄NO₃, 0.3 mM NaH₂PO₄·2H₂O, 0.35 mM K₂SO₄, 1 mM CaCl₂, 1 mM MgSO₄·7H₂O, 20 μM EDTA-Fe, 0.5 mM Na₂SiO₃, 9 μM MnCl₂, 20 μM H₃BO₃, 0.77 μM ZnSO₄, 0.32 μM CuSO₄, and 0.39 μM

(NH₄)₆Mo₇O₂₄, pH 5.5) and grown at either aCO₂ levels or eCO₂ levels for 2 weeks. All nutrient solutions were replaced twice per week, pH was adjusted to 5.5 daily. The average daytime (8:00 a.m. to 8:00 p.m.) CO₂ concentrations were maintained at 598 ± 32 ppm in eCO₂ treatments and 408 ± 28 ppm in aCO₂ treatments. The mean day/night air temperature was maintained at 30 °C/25 °C with 60% relative humidity. The LED lamps were positioned 50 cm above the rice, providing a mean photon flux of 500 μmol m⁻² s⁻¹ during the daytime.

¹⁵N uptake analysis. After growth in hydroponic condition (1.25 mM NH₄NO₃) for 2 weeks, rice root ¹⁵NH₄⁺ and ¹⁵NO₃⁻ influx measurements were performed as described elsewhere^{44,46}. Specifically, 14-day-old rice plants were transferred first to 0.1 mM CaSO₄ for 1 min, then to a complete nutrient solution containing 2.5 mM K¹⁵NO₃ (Sigma, 335134, 98% atom excess ¹⁵N) or 1.25 mM (¹⁵NH₄)₂SO₄ (Aladdin, A110168, 99% atom excess ¹⁵N) instead of 1.25 mM NH₄NO₃ as the N source for 5 min. The plants were incubated in 0.1 mM CaSO₄ for 1 min before the roots were collected and dried at 80 °C for 72 h. Root dry weight was recorded and the ¹⁵N content was measured using the IsoPrime100 elemental analyser (Elementar, Germany). Finally, influx of ¹⁵NH₄⁺ and ¹⁵NO₃⁻ was calculated as described elsewhere²⁴.

Root system analysis. Roots from rice plants grown in the aCO₂ and eCO₂ treatments for 14 d were cut off and spread out in water in a transparent dish. Subsequently, the root system was scanned as described previously²⁴.

Measurement of NR activity. Fresh plant material (-1 g) from individual rice plant grown at either aCO₂ levels or eCO₂ levels for 2 weeks was used to measure NR activity, following the instruction manual of the NR Kit (Solarbio LIFE SCIENCES, BC0080).

Quantitative real time PCR (RT-qPCR) analysis. After growth in hydroponic condition (1.25 mM NH₄NO₃) for 2 weeks, total RNAs were extracted from different plant tissues using the TRIzol reagent (Ambion), and full-length cDNAs were reverse transcribed using a cDNA synthesis kit (Accurate Biology, AG11728). Subsequent RT-qPCR was performed according to the manufacturer's instructions (Accurate Biology, AG11718). Each RT-qPCR assay included at least three biological replicates. As for time course experiment, selected eight time points (0, 15 min, 30 min, 2 h, 24 h, 72 h, 7 d, 14 d) for sample preparation and extracted RNA for RT-qPCR. The rice *ACTIN1* gene was used as an internal reference. Relevant RT-qPCR primer sequences are listed in Supplementary Table 4.

ChIP-qPCR assays. The ChIP-qPCR protocol has been previously described²³. Approximately 2 g of two-week-old *pAct::OsARF6-Flag* and *pAct::OsARF17-Flag* overexpression lines in the Wuyunjing 7 (WYJ7) background, grown under aCO₂ and eCO₂ conditions, were cross-linked with 1% formaldehyde under vacuum for 15 min to stabilize protein-DNA interactions. The samples were then ground to a fine powder in liquid N. Nuclei were isolated and lysed, and chromatin was fragmented by sonication into ~500 bp fragments. The chromatin was incubated overnight at 4 °C with 7 μg of anti-Flag antibody (Sigma, F1804) for immunoprecipitation. The following day, the samples were washed, eluted, and reverse-cross-linked, followed by DNA purification. Enrichment of specific DNA fragments was analyzed by RT-qPCR using three biological replicates. Relevant qPCR primer sequences are provided in Supplementary Table 7.

Western blotting. Total protein was extracted in 50 mM Tris-HCl (pH 7.5), 150 mM NaCl, 0.1% NP-40 detergent, 10% Glycerol, 1 mM DTT with added protease inhibitor cocktail (Roche LifeScience) and transferred onto a nitrocellulose membrane. Then, after blocking with 5% milk solution for 1 h, the nitrocellulose membrane was incubated with

antibodies. The DNRI protein was detected by probing the membrane with anti-DNRI (ABclonal) and proteins involved in photosynthesis were detected by anti-PsbA (Agrisera, AS05084), anti-PsaB (Agrisera, AS10695), anti-RbcS (Agrisera, AS07259), anti-RbcL (Agrisera, AS03037) and anti-SBPase (Agrisera, AS152873), respectively. The result of immunoblotting was visualized on the Tanon-5200 Chemiluminescent Imaging System and grayscale analysis was used Tanon image GIS Semi-quantitative analysis (Tanon Science and Technology).

In vitro transient transactivation assays. As described elsewhere²³, the *indica* variety YD6 was planted under different CO₂ treatments for 10 days. Firstly, the free IAA content of YD6 under two conditions was detected and then extracted rice protoplasts, subsequently, used the effector plasmids *pRTBD-OsARF6* and *pRTBD-OsARF17* to drive the reporters *5×GAL4-OsNRT1.1B*, *-OsNRT2.3a*, *-OsNPF2.4*, *-OsNIA2*, *-OsPsbA2*, *-OsPsbB2*, *-OsRbcS2*, *OsRbcS4*, *OsRbcL* and *OsSBPase*, respectively. Transient transactivation assays were performed as described elsewhere⁴⁷. The Dual-Luciferase Reporter Assay System (Promega, E1960) was used to perform the luciferase activity assays, with the Renilla LUC gene as an internal control. Relevant PCR primer sequences are listed in Supplementary Table 8.

Determination of free IAA content. Root tip samples (~50 mg) were ground into a powder in liquid N and extracted with methanol/water/formic acid (15:4:1, V/V/V). The combined extracts were evaporated to dryness under a N gas stream, reconstituted in 80% methanol (V/V), and filtered through a PTFE membrane (0.22 μm, Anpel). The final solution was then analyzed using an LC-ESI-MS/MS system and an ESI-triple quadrupole-linear ion trap (QTRAP)-MS system (Wuhan Triploid Biotech).

Statistical analysis

We analyzed the data in the field and hydroponic experiments through two-sided Student's *t* tests and two-way ANOVA, including CO₂ and variety as fixed effects. When the ANOVA indicated interactive effects, multiple comparisons were performed using the least significant difference test at the significance level of *P* = 0.05. All analyses were performed with the statistical package SPSS 27.

Reporting summary

Further information on research design is available in the Nature Portfolio Reporting Summary linked to this article.

Data availability

Raw RNA-seq data were deposited at the National Genomics Data Center, Genome Sequence Archive (GSA) (accession number PRJCA024327 [<https://ngdc.cnpc.ac.cn/gsa/s/Hz4L1U30>]). Source data are provided with this paper.

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Author contributions

Y.J. and S.L. designed the study. Y.L., H.Q., and Y.W. conducted the field experiment. S.Z., C.S., and Y.L. conducted the hydroponic experiment. Y.J., S.L., Y.D., Y.L., and S.Z. drafted the paper. Shuijun Hu, W.Z., Shan Huang, S.W., Z.L., G.L., X.F., and K.J.v.G. reviewed and commented on the manuscript.

Competing interests

The authors declare no competing interests.

Additional information

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