



Variation in photosynthetic induction between rice accessions and its potential for improving productivity

Liana G. Acevedo-Siaca¹, Robert Coe^{2,3}, Yu Wang⁴, Johannes Kromdijk^{4,5}, W. Paul Quick^{2,6} and Stephen P. Long^{1,4,7,8}

Department of Crop Sciences, University of Illinois at Urbana-Champaign, Urbana, IL 61801, USA; 2C4 Rice Center, International Rice Research Institute, Los Baños, Laguna 4031, Philippines; ³High Resolution Plant Phenomics Centre, Commonwealth Scientific and Industrial Research Organization (CSIRO), Plant Industry, Canberra, ACT 2601, Australia; ⁴Carl R. Woese Institute for Genomic Biology, University of Illinois at Urbana-Champaign, Urbana, IL 61801, USA; 5Department of Plant Sciences, University of Cambridge, Cambridge, CB2 3EA, UK; ⁶Department of Animal and Plant Sciences, University of Sheffield, Western Bank, Sheffield, S10 2TN, UK; ⁷Department of Plant Biology, University of Illinois at Urbana-Champaign, Urbana, IL 61801, USA; ⁸Lancaster Environment Centre, Lancaster University, Lancaster, LA1 4YQ, UK

Authors for correspondence: Stephen P. Long Tel: +1 217 244 0881 Email: slong@illinois.edu

W. Paul Quick Tel: +63 2 8580 5600 Email: w.p.quick@irri.org

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Summary

- Photosynthetic induction describes the transient increase in leaf CO₂ uptake with an increase in light. During induction, efficiency is lower than at steady state. Under field conditions of fluctuating light, this lower efficiency during induction may cost > 20% of potential crop assimilation. Accelerating induction would boost photosynthetic and resource-use effi-
- Variation between rice accessions and potential for accelerating induction was analysed by gas exchange. Induction during shade to sun transitions of 14 accessions representing five subpopulations from the 3000 Rice Genome Project Panel (3K RGP) was analysed.
- Differences of 109% occurred in the CO₂ fixed during the first 300 s of induction, 117% in the half-time to completion of induction, and 65% in intrinsic water-use efficiency during induction, between the highest and lowest performing accessions. Induction in three accessions with contrasting responses (AUS 278, NCS 771 A and IR64-21) was compared for a range of [CO₂] to analyse limitations. This showed in vivo capacity for carboxylation at Rubisco (V_{c,max}), and not stomata, as the primary limitation to induction, with significant differences between accessions.
- Variation in nonsteady-state efficiency greatly exceeded that at steady state, suggesting a new and more promising opportunity for selection of greater crop photosynthetic efficiency in this key food crop.

Introduction

The efficiency of light interception and conversion to biomass through photosynthesis strongly affects the yield potential of a crop cultivar (Monteith, 1978; Zhu et al., 2010). Understanding of limitations to crop photosynthesis for yield increases have focused on rates of leaf CO2 uptake under conditions of constant light-saturation. However, under field conditions, leaves within a crop canopy experience continually changing light intensity due to intermittent cloud cover, movement of leaves in the wind, and changing solar angles (Walker, 1972; Pearcy, 1990; Burgess et al., 2016). As a result, the photon flux density at any point in the canopy is in continual change, with order of magnitude changes occurring in a matter of seconds (Pearcy, 1990; Pearcy et al., 1996; Zhu et al., 2004; Slattery et al., 2018; De Souza et al., 2019). Dynamic measurements provide insight as to how a plant reacts to the rapid changes in light that occur in crop canopies and can be used to identify limitations to photosynthesis that might be

improved. Transient photosynthetic responses can be categorized by the direction of the light intensity change; that is, from low light to high light or vice versa. This study focuses on the former, photosynthetic induction on shade-to-sun transitions. Induction is considered to be affected by four main processes: photoactivation of enzymes involved in the regeneration and production of ribulose 1,5-bisphosphate (RuBP), build-up of concentrations of the intermediates of carbon (C) metabolism, the activation of Rubisco, and the opening of stomata (Pearcy et al, 1994; Mott & Woodrow, 2000; Slattery et al., 2018; Deans et al., 2019). By definition, photosynthetic efficiency during induction is lower than at steady state, representing a loss of potential crop CO2 uptake (Pearcy et al., 1994; Mott & Woodrow, 2000; Deans et al., 2019). In the case of wheat, this loss over the course of a day was calculated at 21% (Taylor & Long, 2017). If induction could be accelerated, these losses could be reduced and intrinsic water-use efficiency increased (Carmo-Silva et al., 2014; Lawson & Blatt, 2014; McAusland et al., 2016; Vialet-Chabrand et al., 2017).

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Natural genetic variation within crop germplasm forms the basic breeding material to develop new cultivars, and its utilization could aid in improving photosynthetic efficiency, including induction (Flood *et al.*, 2011; Lawson *et al.*, 2012; Driever *et al.*, 2014; Gu *et al.*, 2014; De Souza *et al.*, 2019). Previous experiments have studied photosynthetic induction in soybean and cassava, identifying considerable variation between genotypes (Sakoda *et al.*, 2016; Soleh *et al.*, 2016, 2017; De Souza *et al.*, 2019).

Rice is a direct source of calories for more people than any other single crop. It also serves as the main staple for some 560 million chronically hungry people on the Asian continent (Mohanty, 2013). Improving its photosynthetic efficiency has become a focus, in particular by introducing C₄ photosynthesis (Kajala *et al.*, 2011). Breeding for increased speeds of induction might provide a more immediate and complementary means to increase photosynthetic efficiency (Wang *et al.*, 2020).

Here, the photosynthetic induction of 14 accessions from the 3000 Rice Genome Project (3K RGP) was quantified. These were selected to represent five subpopulations and seven diverse geographical regions. Rubisco activase (Rca) is an important mediator of photosynthetic induction through activation of Rubisco. In rice, it is coded by a single gene that is alternatively spliced to give alpha and beta-isoforms (To *et al.*, 1999). Given the importance of Rca, accessions were also selected on mismatches in the genetic sequence for Rca. The objectives were, first, to compare, quantify and characterize the photosynthetic induction in rice relative to steady state and, second, to examine the response of photosynthetic induction in rice at different [CO₂] to deduce *in vivo* limitations to induction.

Materials and Methods

Germplasm and growing conditions

Fourteen accessions representing five rice subpopulations (*indica*, tropical *japonica*, temperate *japonica*, aus, and aromatic), seven geographical regions, and different canopy structures were taken from the 3K RGP held at the International Rice Research Institute (IRRI) in Los Baños, Philippines (Supporting Information Fig. S1; Table 1). Further selection from single nucleotide polymorphisms (SNPs) was made using the International Rice Informatics Consortium Rice SNP-Seek Database managed by IRRI (Table S2).

Seeds were maintained at 50°C for 1 wk to break dormancy and then sown into soil from the IRRI Upland Farm in small pots (4.5 cm diameter × 12 cm) and fertilized using 0.4 g l⁻¹ of Osmocote Plus 15-9-12 (The Scotts Company Ltd, Thorne, UK). Seedlings were transferred to larger pots (21.5 cm diameter × 21.5 cm, 6 l) after the emergence of the second leaf. These were then placed in a screen house, a type of glasshouse with a glass roof but screen-meshed walls, with no additional lighting or temperature control at IRRI during the Philippines dry season from March to May 2017. Each pot was kept flooded using a drip irrigation system to mimic paddy conditions.

Gas-exchange measurements

Photosynthetic measurements Rice plants were dark adapted overnight. The youngest fully expanded leaf, judged by ligule emergence, was placed in the cuvette of an open gas-exchange system (LI-6400XT; Li-Cor, Lincoln NE, USA). Light was provided by an integrated LED head (2 × 3 LED, LI-6400-02B). Within the cuvette, air temperature was 28°C, flow rate was 400 μmol s⁻¹, [CO₂] was maintained at 400 μmol mol⁻¹, and water vapour pressure deficit at 1-1.6 kPa. For steady-state measurements, leaves were allowed to reach constant rates of CO2 uptake A and stomatal conductance g_s at 1700 μ mol m⁻² s⁻¹ photosynthetic photon flux density (PPFD). For induction, leaves were first allowed to reach a steady state in low light of $50 \, \mu mol \, m^{-2} \, s^{-1}$ PPFD ('shade') for 300 s followed by 720 s at 1700 $\mu mol \, m^{-2} \, s^{-1}$ PPFD ('sun'). Gas-exchange measures were logged every 10 s for the duration of the experiment. Measurements were repeated for all 14 accessions (n=4 plants) in the span of 2 d to minimize any time-dependent effects. Plants were selected at random and measured from 08:00 h to 12:00 h, to avoid confounding accessions with time of day and to minimize any diurnal influences. A, g_s , intercellular CO₂ concentration C_i , transpiration E, and intrinsic water-use efficiency iWUE were calculated following the equations of von Caemmerer and Farquhar (1981). The first 300 s of the induction period were selected to allow a uniform basis for comparing accessions for CO2 uptake and speed of induction. This also represents the period in which most change occurs. However, steady state can take many more minutes to attain, albeit with a small remaining change.

The three accessions that showed the most contrasting induction responses, AUS 278, NCS 711 A and IR64-21, were selected for further analysis of limitations to both induction and steadystate photosynthesis. Determination of the response of A/C_i followed our previously described protocol (Long & Bernacchi, 2003). Induction was measured following the protocol already described for induction, but at a cuvette [CO₂] of either 100, 200, 300, 400, 600 or 800 μmol mol⁻¹ through the induction. The order of cuvette [CO₂] treatments for each individual leaf was randomized to avoid confounding [CO₂] with time. Leaves were dark adapted for a minimum of 1 h between measurements at the different [CO₂]. To determine limitations through induction, A was plotted against C_i for different time points, following the procedure of Soleh et al. (2016). This allowed determination of whether, at any given time point in induction, photosynthesis within the mesophyll was limited by the rate of RuBP regeneration J_{max} , apparent RuBP carboxylase/oxygenase (Rubisco) activity $V_{c, max}$, and to quantitatively partition nonstomatal and stomatal limitations at each time point during induction, following the methods of Bernacchi et al. (2003) and Kaiser et al. (2017), as outlined in the following.

Calculations Nonstomatal and stomatal limitations were calculated using the following equations, where the subscript 'time' refers to the minimum limitation at a given point in time during induction and the subscript 'final' refers to the minimum limitation at the final time point during induction. These equations

Table 1 Minimum, maximum, mean, percentage genetic variation (PGV), and significance of variation assessed by ANOVA for each steady-state and non-steady-state measure across 14 rice (*Oryza sativa*) accessions.

Trait	Unit	Min.	Max.	Mean	PGV (%)	df	P-value	Р
Dynamic traits								
\overline{A}_{300}	μ mol m ⁻² s ⁻¹	5.8	18	12	101.6	13	< 0.0001	* * *
A ₇₀₀	μ mol m ⁻² s ⁻¹	9.79	23.81	19.75	71	13	< 0.0001	* * *
gs avg	$mol m^{-2} s^{-1}$	0.1	0.71	0.33	189.6	13	< 0.0001	* * *
iWUE _{avg}	μ mol CO ₂ mol ⁻¹ H ₂ O	22.6	71.01	40.13	120.9	13	0.026	*
C_{iavg}	μmol mol ⁻¹	260.3	341.3	310.3	26.1	13	0.07	
A ₃₀₀	μ mol m ⁻² s ⁻¹	10.3	24	16.8	81.7	13	< 0.0001	* * *
A_{Max}	$_{\mu}^{\mathrm{mol}}\mathrm{m}^{-2}\mathrm{s}^{-1}$	11.8	32.9	21.5	98.1	13	< 0.0001	* * *
IT ₉₀	S	27.3	127.5	65.9	152	13	0.0014	* *
IT ₉₀	S	99	289.6	201.9	94	13	0.216	
$IT_{90} - IT_{50}$	S	54.18	229.6	136	129	13	0.519	
C Loss ₃₀₀	μ mol m ⁻² s ⁻¹	807	8298	4281	175	13	0.001	* *
C Loss ₇₀₀	μ mol m ⁻² s ⁻¹	-1260	16 737	7105	253.3	13	0.003	* *
Steady-state tra	aits							
Α	μ mol m ⁻² s ⁻¹	19.5	35.1	26.4	59.1	13	0.035	*
gs	$mol m^{-2} s^{-1}$	0.26	1.45	0.72	166.1	13	0.161	
Ci	μmol mol ⁻¹	240.7	328	289	30.2	13	0.191	
Ε	$^{\circ}$ mmol m ⁻² s ⁻¹	4.2	12.9	8.5	102.3	13	0.72	
iWUE	μ mol CO ₂ mol ⁻¹ H ₂ O	20.1	76.3	40.3	139.3	13	0.174	

Traits are as follows: \overline{A}_{300} and \overline{A}_{700} , average A during the first 300 s and 700 s of induction, respectively; g_s $_{avg}$, average stomatal conductance during the first 300 s of induction; iWUE $_{avg}$, average intrinsic water-use efficiency (iWUE $=\overline{A}_{300}/g_{savg}$); C_{i} $_{avg}$, average intercellular [CO $_2$] during the first 300 s; A_{300} , A at 300 s into induction; A_{Max} , maximum rate of CO $_2$ uptake across the entire induction period; IT $_{50}$ and IT $_{90}$, time that A reached 50% and 90%, respectively, of A_{300} ; IT $_{90}-IT_{50}$, the difference between IT $_{90}$ and IT $_{50}$; C Loss $_{300}=(A-\overline{A}_{300})\times 300$, the difference between the total uptake that would have occurred over the first 300 s if A had risen instantaneously to A_{300} less the integral of the actual A over the first 300 s; C Loss $_{700}=C$ Loss $_{700}=(A-\overline{A}_{700})\times 700$; A_{sat} , light-saturated leaf CO $_2$ uptake; g_s , stomatal conductance; C_i , intercellular [CO $_2$] at steady state; intrinsic water-use efficiency (iWUE $=A/g_s$); PGV= (Min. -Max.)/Mean. *, P<0.05; ***, P<0.01; ***, P<0.01; ***, P<0.001; degrees of freedom.

were adapted from those of Kaiser *et al.* (2017). In the absence of diffusional limitation, the CO_2 uptake rate at a given point in induction would be A_C^* :

$$A_{C_a}^* = A \times \frac{\min\{A_c(C_a), A_j(C_a), A_t(C_a)\}_{time}}{\min\{A_c(C_i), A_j(C_i), A_t(C_i)\}_{final}}$$
 Eqn 1

$$A_{C_{i}}^{*} = A \times \frac{\min\{A_{c}(C_{i}), A_{j}(C_{i}), A_{t}(C_{i})\}_{\text{time}}}{\min\{A_{c}(C_{i}), A_{j}(C_{i}), A_{t}(C_{i})\}_{\text{final}}}$$
Eqn 2

(A, the CO₂ uptake rate at any given point in the induction; A_c , A_j and A_t are the rates that can be supported at that time point and a given [CO₂] by Rubisco activity, RuBP regeneration, and triose-phosphate utilization, respectively; C_a , the chamber [CO₂]; C_i , intercellular [CO₂]).

Nonstomatal limitation L_{NS} and stomatal limitation L_{S} were then calculated:

$$L_{\text{Nonstomatal}} = \frac{A_{\text{f}} - A_{\text{G}}^*}{A_{\text{f}} - A_{\text{i}}} \times 100$$
 Eqn 3

$$L_{\text{Stomatal}} = \frac{A_{C_a}^* - A}{A_f - A_i} \times 100$$
 Eqn 4

($A_{\rm f}$, equal to the final or steady-state CO₂ uptake rate at the end of induction; $A_{\rm i}$, the value prior to induction during shade). All calculated limitations are relative to the final value of the induction. Since limitations are based on $C_{\rm i}$, and not chloroplast

[CO₂], mesophyll conductance will be included in the nonstomatal limitations.

C Loss, the integrated amount of CO₂ uptake lost due to the lower rates through induction compared with steady state, was calculated thus:

$$CLoss_t = (A - \overline{A}_t) \times t$$
 Eqn 5

(A, the steady-state rate of uptake; \overline{A}_t , the average rate across the measured time period from the start of the induction t, either 300 s or 700 s).

Percentage genetic variation (PGV) was calculated thus:

$$PGV = \frac{X_{\text{max}} - X_{\text{min}}}{\overline{X}} \times 100$$
 Eqn 6

 $(X_{\text{max}}, X_{\text{min}} \text{ and } \overline{X})$, the maximum, minimum and mean values, respectively, of each trait across the 14 accessions).

Statistical analyses

All statistical analyses and model-fitting used R (v.3.5.2). Normal distribution and homogeneity of variances were tested by the Shapiro–Wilk test and Brown–Forsythe test, respectively. For data conforming to both assumptions, ANOVA was performed followed by Tukey's mean discrimination analysis, using the R package AGRICOLAE. Correlations between measured parameters were assessed using the Pearson correlation analysis (R packages

CORRPLOT and HMISC). Accession means were used for the Pearson correlation analysis.

Results

Photosynthetic induction responses vary significantly between 14 rice accessions

Induction of CO_2 uptake and stomatal conductance showed biphasic responses to a change from 50 to 1700 μ mol m⁻² s⁻¹ PPFD. During the first 120 s of induction, A and g_s both increased rapidly, followed by a more gradual increase (Fig. 1). At steady state, the only trait that varied significantly between accessions was A (P=0.035) (Table 1), whereas almost all traits showed significant variation under the nonsteady-state conditions of induction (Fig. 1; Table 1). Nonsteady-state measurements showed c. 20–40% greater variation between accessions relative to the equivalent trait at steady state (Table 1).

Averaged across the induction period, CO_2 uptake \overline{A}_{300} , stomatal conductance $g_{s \text{ avg}}$, intrinsic water-use efficiency during induction iWUE_{avg}, maximum CO_2 uptake during induction A_{Max} , and the time to 50% induction in seconds IT₅₀ all varied significantly between accessions (Table 1). The substantial range of variation was evident in PGV: 102% for \overline{A}_{300} , 190% for $g_{s \text{ avg}}$, 121% for iWUE_{avg}, and 152% for IT₅₀ (Table 1). There were no significant differences in $C_{i \text{ avg}}$, the average C_{i} over the induction (Table 1). The highest and lowest performing accession means differed 109% with respect to \overline{A}_{300} ($P \le 0.0001$) (Figs 1, 2).

These differences between accessions were independent of subpopulation, geographic region, or canopy structure (Figs 2, S1, S2; Table S1). The accession mean that had the highest \overline{A}_{300} was Malogbana (16.8 μmol m⁻² s⁻¹), an admixed accession from d'Ivoire, whereas the lowest was AUS 278 (7.6 µmol m⁻² s⁻¹), an aus accession from Bangladesh (Fig. 2; Table S1). There was a 65% difference in iWUE_{avg}, between accession means over the first 300 s of induction (50.2-30.9 µmol CO₂ mol⁻¹ H₂O; Fig. 2). Additionally, there was a significant 117% difference in IT₅₀ between accessions (Fig. 2). The accession showing the slowest induction, as determined by IT₅₀, was Du Gen Chuan, a Chinese indica accession (101 s), whereas the fastest was JC1, a Chinese aromatic accession (43 s; Fig. 2; Table S1). Despite significant differences in IT₅₀, the time to 90% induction IT₉₀ did not vary significantly between accessions, indicating that most of the variation occurs in the early phase of induction (Fig. 2; Table 1). Loss of potential C fixation due to the lag that occurs in photosynthesis through the first 300 s of induction (C Loss₃₀₀) and 700 s (C Loss₇₀₀) varied significantly between accessions (Table 1).

No significant relationship between steady-state and induction measures of photosynthesis

There was no significant correlation between \overline{A}_{300} or the speed of induction and A at steady state and, indeed, no significant correlation between any induction trait and its steady-state equivalent (Fig. 3). Correlations were found between different measures

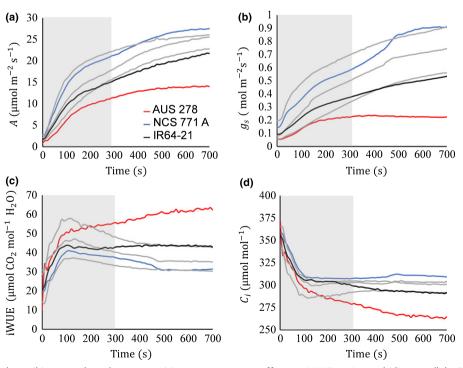


Fig. 1 (a) Net leaf CO₂ uptake A, (b) stomatal conductance g_{sr} (c) intrinsic water use efficiency iWUE = A/g_{sr} and (d) intercellular [CO₂] C_i with time t of induction on transfer at 0 s from low to high light (50–1700 μ mol m⁻² s⁻¹) in rice ($Oryza\ sativa$). The 'induction period', characterized by a steep increase in A, is demarked with the grey box and lasted up to 300 s. The three accessions that were selected for further study of induction at varied [CO₂] were AUS 278 (red), NCS 771 A (blue) and IR64-21 (black). The other three accessions in grey are Du Gen Chuan, K2 C45 and Malogbana. For ease of visualization, only six accessions are included, but the data for all 14 accessions are available in Table 1 and Fig. 2.

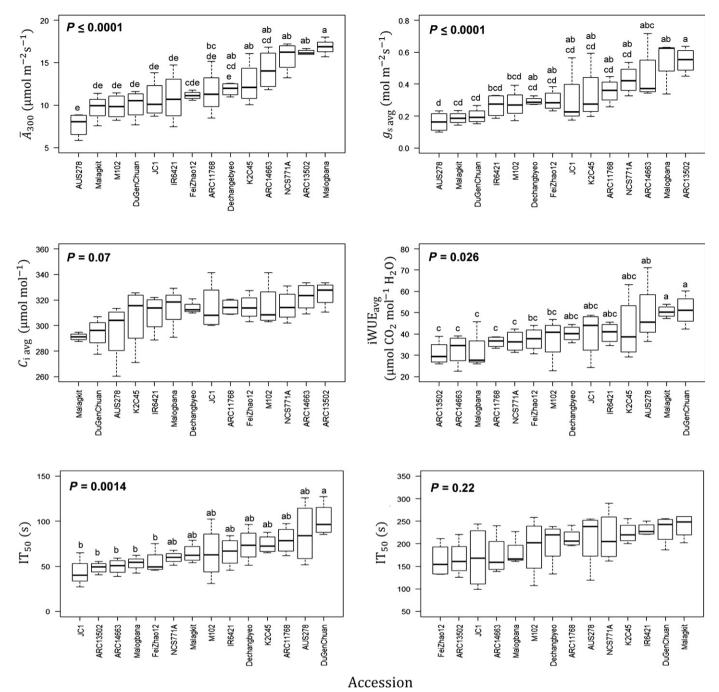


Fig. 2 Mean and variation for all rice (*Oryza sativa*) accessions for average leaf CO_2 uptake during the first 300 s of induction (\overline{A}_{300}), average A stomatal conductance during the first 300 s of induction ($g_{s \text{ avg}}$), average intrinsic water-use efficiency (iWUE_{avg} = $\overline{A}_{300}/g_{savg}$), average intercellular CO_2 concentration during the first 300 s ($C_{i \text{ avg}}$), and time that A reached 50% and 90% of A_{300} (IT₅₀ and IT₉₀, respectively). The accessions are ranked by increasing mean for each parameter. Different letters represent statistically significant differences (P < 0.05) between different accessions.

made within light induction and within steady state, but not between. Among induction traits, significant positive correlations were found between \overline{A}_{300} and $g_{s \text{ avg}}$, \overline{A}_{300} and A_{Max} , and \overline{A}_{300} and \overline{A}_{660} (Fig. 3). Significant negative correlations were found between \overline{A}_{300} and iWUE_{avg}, iWUE_{avg} and $g_{s \text{ avg}}$, and \overline{A}_{300} and CLoss₃₀₀ (Fig. 3). Interestingly, a significant negative correlation was found between \overline{A}_{300} and speed of induction, with plants that had a lower IT₅₀ and IT₉₀ assimilating more CO₂ (Fig. 3b,d). Additionally, faster stomatal opening and greater g_s at the

beginning of induction were significantly correlated with greater \overline{A}_{300} and quicker IT₅₀ and IT₉₀ (Figs S3, S4). Furthermore, plants responded consistently for g_s throughout induction, as indicated by a strong correlation (P= 0.0085); that is, a high g_s early in induction was consistent with a high g_s at the end of induction (Fig. S4). For measures made at steady-state, positive correlations were predictably found between g_s and G_s and between g_s and G_s and negative correlations between iWUE and G_s and between iWUE and G_s and between iWUE and G_s (Fig. 3).

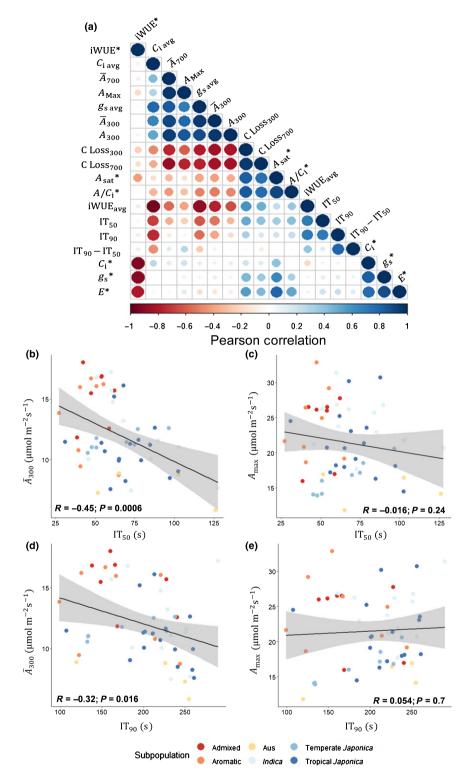


Fig. 3 (a) Pearson correlation R^2 of all measured dynamic and steady-state (*) photosynthetic traits measured in rice ($Oryza\ sativa$). Negative correlations (red) and positive (blue). Traits are as follows: \overline{A}_{300} and \overline{A}_{700} , average A during the first 300 s of induction, respectively; A_{sat} , light-saturated leaf CO_2 uptake; g_s , stomatal conductance; C_i , intercellular $[CO_2]$ at steady state; iWUE, intrinsic water-use efficiency (iWUE = A/g_s); $g_s\ av_g$, average stomatal conductance during the first 300 s of induction; iWUE $_{avg}$, average intrinsic water-use efficiency (iWUE = $\overline{A}_{300}/g_{savg}$); $C_i\ av_g$, average intercellular $[CO_2]$ during the first 300 s; A_{300} , A at 300 s into induction; A_{Max} , maximum rate of CO_2 uptake across the entire induction period; IT_{50} and IT_{90} , time that A reached 50% and 90%, respectively, of A_{300} ; $IT_{90} - IT_{50}$, the difference between IT_{90} and IT_{50} ; C Loss $_{300} = (A - \overline{A}_{300}) \times 300$, the difference between the total uptake that would have occurred over the first 300 , if A had risen instantaneously to A_{300} less the integral of the actual A over the first 300 s; C Loss $_{700} = (A - \overline{A}_{700}) \times 700$. (b–e) Individual measures, regression line, correlation coefficient R^2 , confidence interval (95%), and R-value for (b) \overline{A}_{300} vs IT_{50} , (c) \overline{A}_{300} vs IT_{90} , (d) A_{Max} vs IT_{90} , and (e) A_{Max} vs IT_{90} .

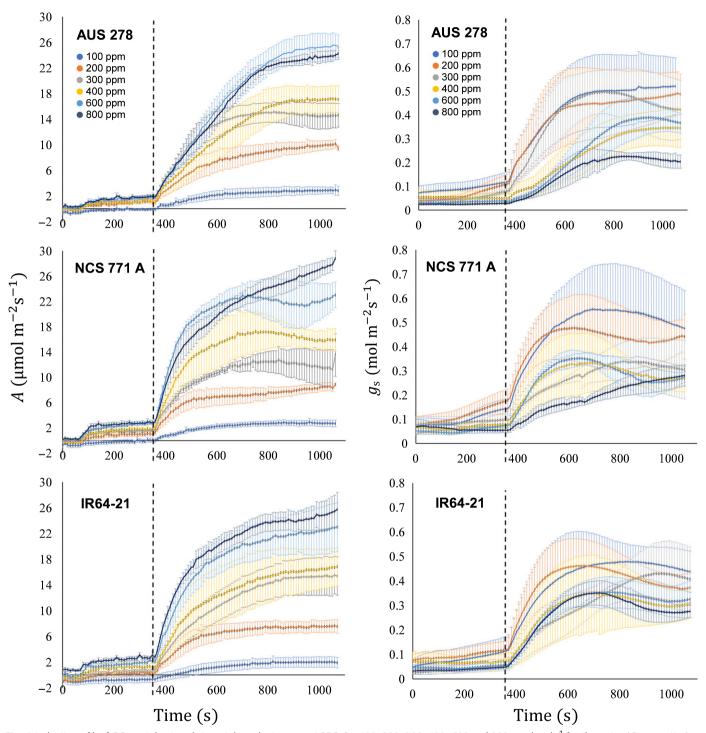


Fig. 4 Induction of leaf CO₂ uptake A and stomatal conductance g_s at [CO₂] = 100, 200, 300, 400, 600 and 800 μ mol mol⁻¹ for three rice (*Oryza sativa*) cultivars. Each point is the mean (\pm SE) of four plants. The dashed vertical line at 360 s represents the point of transition from low to high photosynthetic photon flux density (50–1700 μ mol m⁻² s⁻¹); that is, the start of induction.

Rice photosynthetic induction is mainly limited by biochemistry

Three accessions were selected, based upon the variation in their \overline{A}_{300} : NCS 771 A, IR64-21 and AUS 278; 15.7 m⁻² s⁻¹, 10.9 m⁻² s⁻¹ and 7.7 m⁻² s⁻¹, respectively. Induction was

measured on the three selected accessions at six $[CO_2]$ (Fig. 4). Dynamic A/C_i curves were constructed at four time points: 60, 180, 360 and 700 s after the beginning of induction (Fig. 5). The temporal changes in the dynamic A/C_i curves were highly dependent upon the accession (Fig. 5). When compared, dynamic curves for NCS 771 A strongly approximated the steady-state A/C_i

 C_i curves at both 360 s and 700 s. Dynamic A/C_i curves in AUS 278 were only similar to the steady-state curve at 700 s, whereas A values for IR64-21 were lower than at steady state at all C_i s, even at 700 s (Fig. 5).

When stomatal and nonstomatal limitations were calculated relative to those at steady state for all three accessions, nonstomatal limitation $L_{\rm NS}$ accounted for almost 100% initially, declining to 30% at 120 s and then to no more limiting than at steady state (Fig. 6). In contrast, stomatal limitation $L_{\rm S}$ gradually increased from c. 2% to c. 10–15% over the first 300 s (Fig. 6). Similar patterns were seen in IR64-21, AUS 278 and NCS 771 A, although $L_{\rm NS}$ decreased more rapidly in NCS 771 A. Consistent with strong $L_{\rm NS}$, $C_{\rm i}$ avg was higher during induction than $C_{\rm i}$ at steady state (Table 1). This is reflected in the dynamic $A/C_{\rm i}$ responses, where $C_{\rm i}/C_{\rm a}$ was higher than at steady state at all [CO₂], again indicating that stomatal conductance was less of a limitation during induction than at steady state (Fig. 6). Since estimates of $L_{\rm NS}$ and $L_{\rm S}$ were based on $C_{\rm i}$, any limitation due to mesophyll conductance is included in $L_{\rm NS}$.

The dynamic A/C_i curves suggested that photosynthetic induction was strongly limited by Rubisco (Fig. 5). Fitting $V_{c,max}$ to the A/C_i responses showed its sharp increase during the first 300 s of induction, followed by a more gradual increase to steady-state values (Fig. 7). $V_{c,max}$ during induction differed significantly between the three accessions (Fig. 7).

Discussion

Possibilities for exploiting natural variation in the photosynthetic induction response of rice

Natural genetic variation is one of the key drivers of evolution and is essential for plant improvement through breeding. Although breeding selection for higher productivity has resulted in only very small increases in leaf photosynthetic capacity, recent studies have suggested variation within crop diversity panels could be exploited for selection based on direct measurement of photosynthesis (Flood et al., 2011; Lawson et al., 2012; Driever et al., 2014; Gu et al., 2014; Koester et al., 2016). However, these studies have only examined steady-state photosynthesis. Given that photosynthesis is probably never at steady state in crop fields, the speed of adjustment to dynamic lighting may be as or more important. Recent studies have suggested that speeding up the rate of adjustment to sunlight-to-shade and shade-to-sunlight transitions could each increase canopy CO₂ uptake in the field, each by c. 20% (Kromdijk et al., 2016; Taylor & Long, 2017).

Here, gas-exchange methods were used to determine the potential extent of diversity and identify useful parameters. Though only a few genotypes could be examined by these methods, they represent a broad cross-section of rice germplasm (Table S1). Significant natural variation was found between accessions for the nonsteady-state measures \overline{A}_{300} , IT_{50} and $\mathrm{iWUE}_{\mathrm{avg}}$, indicating variation in rice that could be exploited to improve photosynthesis in dynamic conditions. Improving the

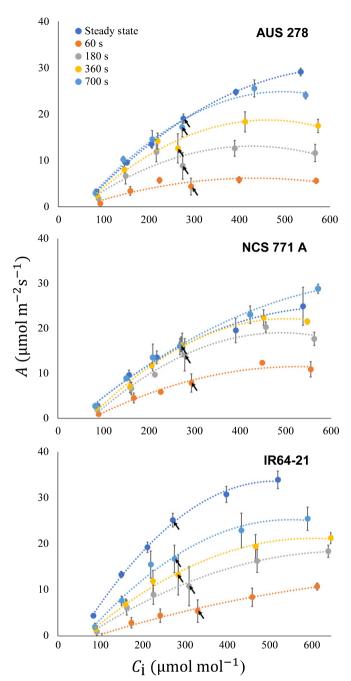


Fig. 5 The responses of leaf CO₂ uptake A to intercellular [CO₂] C_i at steady state (dark blue), and at 60 s (orange), 180 s (grey), 360 s (yellow), and 700 s (light blue) from the start of induction and replotted from Fig. 4. The operating point of each curve at 400 μ mol mol⁻¹ atmospheric [CO₂] C_a is indicated with a black arrow. Each point is the mean (\pm SE) of four plants of each rice (*Oryza sativa*) accession.

speed of photosynthetic induction would be particularly desirable as it would allow the plant to respond more rapidly to fluctuations in its light environment, capitalizing on available photosynthetically active radiation (Taylor & Long, 2017). These improvements in photosynthetic induction could then translate to increases in plant productivity, especially if combined with increased overall photosynthetic capacity (Taylor & Long, 2017). Indeed, IT $_{50}$ was strongly correlated with \overline{A}_{300} , suggesting that

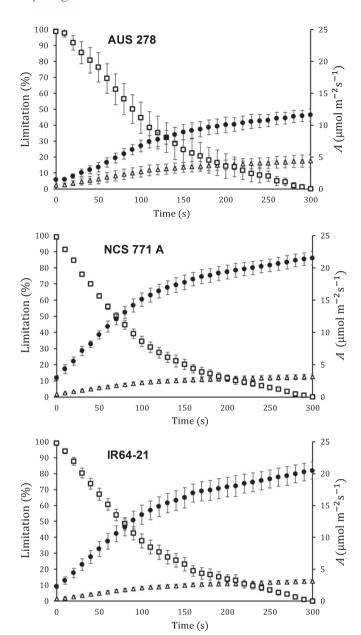


Fig. 6 Nonstomatal (open squares) and stomatal (closed triangles) limitations with leaf CO_2 uptake A (closed circles) vs time over the first 300 s of induction for the three selected rice (*Oryza sativa*) accessions. Nonstomatal and stomatal limitations during induction were calculated relative to the near-steady-state value obtained at 300 s. Each point represents the mean (\pm SE) of four individual plants measured at ambient [CO_2].

both parameters could be improved simultaneously (Fig. 3). However, as in steady-state conditions, trade-offs exist between photosynthetic performance and water-use efficiency, which need to be taken into consideration. Undoubtedly, breeding for increased photosynthetic efficiency in fluctuating light would require screening and analysis of more genotypes than examined here. Modulated Chl fluorescence imaging would be an effective method to screen large diversity collections and the progeny of any crosses. The current study has shown biochemical limitations, rather than stomatal limitations, as the major factor

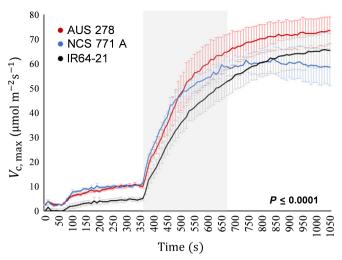


Fig. 7 Apparent maximum rate of carboxylation $V_{c,max}$ with time through induction calculated from the response of leaf CO_2 uptake A to intercellular $[CO_2]$ C_i derived from inductions made at different $[CO_2]$, as in Fig. 4. Induction began at 360 s. Significance of difference between the three rice ($Oryza\ sativa$) accessions is based on a repeated measures (time) one-way ANOVA. Each point represents the mean (\pm SE) of four individual plants.

influencing CO₂ uptake during the first few minutes of a shade to sun transition in rice. Biochemical limitation could be screened through fluorescence imaging by monitoring the increase in efficiency in photosystem II through induction of leaf disks in 96-well plates, for example. This would allow the screening of several hundred genotypes in a day (Murchie & Lawson, 2013).

There was no obvious association of induction parameters with geographic region or genetic grouping, suggesting wide crosses would not be needed to obtain variation in these parameters for selection. Here, we also found that crop canopy (Fig. S1) did not correlate with photosynthetic performance during induction (Fig. S2). Despite their contrasting photosynthetic induction performances, both NCS 771 A and AUS 278 had drooping canopies, whereas IR64-21 had an erect canopy structure (Fig. S1). Additionally, it was found that IR64, an elite high-yielding variety released in 1985, was significantly outperformed by several accessions for both dynamic and steady-state photosynthetic traits (Fig. 1). At the height of its popularity, IR64 was farmed on over 10×10^6 ha world-wide while accounting for 70% of rice area planted in Indonesia and 10% of breeder seed produced in India (Mackill & Khush, 2018). IR64 is still widely cultivated across much of tropical Southeast and South Asia, although it has been replaced by newer varieties, many of which are its progeny or relatives (Mackill & Khush, 2018). IR64 was bred by initially selecting for semi-dwarf morphology, increasing harvest index but not photosynthetic performance (Mackill & Khush, 2018). This study underscores the importance of screening landraces and accessions that are not widely cultivated, as they may perform better than elite varieties, with respect to photosynthesis in fluctuating light, and act as a source for plant improvement.

There is also untapped potential to improve photosynthetic induction by screening wild rice species and the progenitors of rice (*Oryza rufipogon* and *Oryza nivara*) which can exhibit greater rates for leaf CO₂ uptake when compared to cultivated *Oryza sativa* accessions (Zhao *et al.*, 2008).

How [CO₂] influences the response of photosynthetic induction

The response of photosynthetic induction to elevated [CO₂] is an important consideration given the current rapid rise in global atmospheric [CO₂] (IPCC, 2018). Previous in-field studies that have focused almost exclusively on steady-state photosynthesis show a large increase in CO₂ uptake of 25-50% almost in proportion to the increase in [CO₂]. However, yield increases measured under nonsteady-state open-air field elevation of [CO₂] in rice are considerably less than the increases predicted from controlled environment studies (Long et al., 2006; Ainsworth, 2008; Cai et al., 2015). Here, we found that rice saw increased leaf CO_2 uptake in response to elevated $[CO_2]$. However, \overline{A}_{300} increased on average by 43% between 400 and 800 µmol mol⁻¹, whereas steady-state A increased by 53% (Fig. 4), showing that the increases are smaller in fluctuating light than at steady state. This could be part of the explanation as to why productivity gains in field conditions are smaller than those observed under elevated [CO₂] in controlled environments.

Photosynthetic induction in rice is limited mostly by biochemistry

Here, C_i was consistently higher during induction despite having lower A, indicating a biochemical rather than stomatal limitation to photosynthesis (Table 1). $L_{\rm NS}$ was found to be more limiting than $L_{\rm S}$, except in the final phase of induction when $L_{\rm S}$ became more prominent (Fig. 5). Analysis of the A/C_i response indicated $V_{\rm c,max}$ to be limiting throughout induction in all three accessions (Fig. 5), similar to previous results across a wide range of genotypes in soybean (Soleh *et al.*, 2017). As noted in the Results section and as in the Soleh *et al.* (2017) study, our methods do not separate limitations due to mesophyll conductance from biochemical limitations. Changes in mesophyll conductance have been suggested to be far more rapid than induction of carboxylation, but this clearly requires further investigation (Deans *et al.*, 2019).

Limitations to photosynthetic induction appear to be species dependent. Soy and wheat are primarily limited by their biochemistry, whereas cassava and some tropical tree species are more heavily limited by stomata (Tinoco-Ojanguren & Pearcy, 1993; Valladares *et al.*, 1997; Soleh *et al.*, 2016; Taylor & Long, 2017; de Souza, *et al.*, 2019). Species that have a weaker coupling between A and gs responses in dynamic light conditions have greater limitation from stomata and exhibit a stronger lag in gs in reaching steady state (McAusland *et al.*, 2016). In the present study, stomata only accounted for c. 10% of limitation to A during photosynthetic induction (Fig. 5), similar to that found for rice by McAusland *et al.*, 2016. The dumb-bell-shaped guard

cells in rice may contribute to a faster stomatal response as they require fewer solutes and less water for a change in aperture, allowing them to respond more quickly than elliptical-shaped guard cells (McAusland *et al.*, 2016). Additionally, plants with smaller stomata, such as rice, are known to respond more quickly to environmental stimuli, which may reduce limitation by stomata in dynamic conditions (Ohsumi *et al.*, 2007; Drake *et al.*., 2013; Raven, 2014). It is possible that L_S may not play such a strong role in rice induction. Its wild ancestors were emergent aquatic plants, and most breeding programmes that target the improvement of lowland rice utilize flooded paddies, where water is not limiting. After millennia of cultivating rice in conditions with plentiful water, it is likely that rice is less conservative in its water usage, explaining the lower L_S (Nay-Htoon *et al.*, 2018).

The lack of correlation between dynamic and steady-state photosynthesis in rice

Here, no significant relationship was found between steady-state and the equivalent nonsteady-state trait during induction (Fig. 2), paralleling prior work with soybean (Soleh *et al.*, 2016). Accessions that had high CO_2 uptake during induction or a speedier induction rate did not necessarily have higher CO_2 uptake in steady-state measurements (Fig. 2). For example, rates of induction that appear dominated by Rubisco activation are not correlated with steady-state $V_{\mathrm{c,max}}$, suggesting a lack of correlation between Rca activity and Rubisco activity at steady state.

The lack of correlation between the steady-state photosynthetic phenotype and induction phenotypes challenges the way that photosynthesis has conventionally been measured and understood for selection in the field environment. In the absence of significant correlations between dynamic and steady-state photosynthetic phenotypes, therefore, criteria used in the selection of photosynthetic efficiency in crop improvement of productivity need to be rethought. The results here indicate that consideration of steady-state criteria alone has likely failed to account for the larger part of phenotypic diversity in crop germplasm with respect to photosynthesis. Variation across the germplasm examined in photosynthetic traits during induction was 40% greater than in their steady-state counterparts (Table 1). Though improving photosynthesis remains a major opportunity for improving genetic yield potential (Evans, 1997; Zhu et al., 2008, 2010; Long et al., 2015), it appears that more focus on nonsteady-state traits is needed.

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Author contributions

LGA-S and SPL planned the research; SPL and WPQ supervised the project. LGA-S and RC conducted the experimental work, LGA-S analysed the data, LGA-S and JK conducted limitations analysis, YW conducted the modelling, and LGA-S and SPL wrote the manuscript with the input of all the other authors.

ORCID

Liana G. Acevedo-Siaca https://orcid.org/0000-0003-3903-0402

Robert Coe https://orcid.org/0000-0002-4832-4376
Johannes Kromdijk https://orcid.org/0000-0003-4423-4100
Stephen P. Long https://orcid.org/0000-0002-8501-7164
Yu Wang https://orcid.org/0000-0002-6951-2835

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Supporting Information

Additional Supporting Information may be found online in the Supporting Information section at the end of the article.

- **Fig. S1** 14 selected rice accessions during mid-tillering to show differences in canopy.
- **Fig. S2** Relationship between average CO_2 uptake during induction (\overline{A}_{300}) and leaf angle, the time to 50% induction (IT₅₀) and leaf angle.
- **Fig. S3** Correlations between average CO_2 uptake during induction (\overline{A}_{300}) and time to 50% stomatal opening (g_{s}) , time to 90% stomatal opening (g_{s}) , time to 50% induction (IT_{50}) , and time to 90% induction (IT_{90}) .
- **Fig. S4** Correlations between initial and final stomatal conductance during induction and speed of induction.
- Table S1 Description of selected accessions.
- **Table S2** SNP sequences for Rubisco activase for the 14 selected accessions.

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