

1 **Phenotyping Photosynthesis on the Limit – A critical examination of RACiR.**

2 We are now in the phenomic era. New, rapid screening approaches that provide quality data
3 on plant physiological performance are a priority. A recently published advance, enabled by a
4 next generation gas exchange system (LI-6800, LI-COR, Lincoln NE, USA), is the Rapid A-
5 Ci Response (RACiR: Stinziano et al., 2017). This technique aims to allow rapid (~12 min)
6 determination of the A/c_i response (net CO₂ assimilation (A) response to intercellular CO₂
7 concentration (c_i)), which previously required ~40 minutes (standard A/c_i approach).
8 Evaluating RACiR as a means for higher throughput phenotyping reveals offsets to
9 commonly estimated parameters from A/c_i analyses; however, best practices for RACiR
10 application and reporting should enable reliable comparisons with standard A/c_i
11 measurements.

12 The A/c_i response is a powerful tool used by plant physiologists to establish factors
13 limiting photosynthesis *in vivo* and *in situ*. In combination with biophysical models, A/c_i
14 measurements can establish apparent rates for Rubisco carboxylation (V_{cmax}), electron
15 transport capacity (J), and, triose-phosphate utilisation (TPU), as limits affecting
16 photosynthesis (Long & Bernacchi, 2003). The term apparent is used here, since the true
17 values of the parameters requires the response of A/c_c , where c_c is [CO₂] within the
18 chloroplast, which can be determined by simultaneous measurement of mesophyll
19 conductance (g_m) (Bernacchi *et al.* 2002). These parameters are used in applications from
20 crop improvement (Kromdijk & Long, 2016) to global change modelling (Rogers *et al.*,
21 2014). An important application of A/c_i response measurements is in understanding
22 regulation of plant carbon and water balance by partitioning limits to photosynthesis into
23 diffusive and biochemical components (Farquhar & Sharkey, 1982; Jones, 1985; Buckley &
24 Diaz-Espejo, 2015). In addition, by establishing the $c_{i,trans}$ at which electron transport or TPU
25 limitation replace carboxylation as limiting for photosynthesis, trends in c_i under operating
26 conditions can be used to identify limiting states under dynamic conditions (e.g., Taylor &
27 Long, 2017).

28 Measurements of A/c_i responses using the RACiR rely on comparisons of sample
29 (CO_{2,S}) and reference (CO_{2,R}) [CO₂] under a continuous, linear ramp in CO_{2,R}; possible
30 because of the LI-6800 photosynthesis system. Following Stinziano et al., (2017) the linear
31 relationship between sample and reference IRGA [CO₂] from a measurement with an empty
32 chamber is used to correct A .

33 $A_{EMPTY} = m \cdot CO_{2,R} + c$

34 $A_{CORR} = A_{MEAS} |CO_{2,R} - A_{EMPTY}| CO_{2,R}$

35 Where A_{EMPTY} is the net CO_2 assimilation rate measured with the empty chamber; m and c
 36 are a slope and intercept respectively; A_{CORR} is the corrected, A_{MEAS} the measured, and
 37 A_{EMPTY} the apparent empty chamber net CO_2 assimilation rate. In addition, c_i must be
 38 corrected to match A_{CORR} . Substituting A_{CORR} for A in the manufacturer's equation:

39
$$c_{i,CORR} = \frac{\left(g_{tc} - \frac{E}{2}\right) CO_{2,R} - A_{CORR}}{\left(g_{tc} + \frac{E}{2}\right)}$$

40 Where g_{tc} is the total leaf conductance to CO_2 , and E is the rate of transpiration. These
 41 corrections to A and c_i attempt to integrate correction of errors resulting from sources with
 42 different time constants: volume differences of the sample and reference flow paths, impacts
 43 of the particular flow rate used, physical distance between CO_2 injector and reference
 44 chamber, and cumulative calibration offsets (IRGA matching; Stinziano et al., 2017).

45 Experimental results suggest that RACiR measurements demonstrate a consistent
 46 offset towards lower CO_2 compensation points (Γ) (Fig. 1). This was tested using both a C_3
 47 species (cowpea, *Vigna unguiculata*; Fig. 1a) and a species with C_4 photosynthesis, teff
 48 (*Eragrostis teff*), which we expected to show a compensation point close to zero. We found
 49 that the RACiR suggested an impossible, negative compensation point for teff (Fig. 1b).
 50 Returning to the data published by Stinziano et al., (2017) we also found unrealistic offsets in
 51 Γ (Fig. 1c).

52 To establish the importance of these offsets, the Farquhar et al., (1980) model (FvCB
 53 model) was fit to RACiR data of Stinziano et al., (2017) using 500-0 $\mu mol mol^{-1}$ and 300-800
 54 $\mu mol mol^{-1}$ ramps in $CO_{2,R}$, compared with overlapping data for standard A/c_i measurements.

55
$$A = \min(A_C, A_J) \left(1 - \frac{\Gamma^*}{c_i}\right) - R_d$$

56
$$A_C = V_{c,max} \frac{c_i}{c_i + K_{CO}}$$

57
$$A_J = J \frac{c_i}{4.5c_i + 10.5\Gamma^*}$$

58 A_C is Rubisco-limited, A_J electron transport-limited CO_2 assimilation, and R_d dark
59 respiration. The parameters Γ^* (photorespiratory compensation point) and K_{CO}
60 ($K_{CO}=K_C(1+O/K_O)$: where K_C and K_O are temperature dependent, Rubisco Michaelis-
61 Menten constants for CO_2 and O_2 ; and O is atmospheric oxygen concentration) were
62 obtained from Bernacchi et al., (2001). Estimating $V_{c,max}$, J and R_d from the data, led to
63 similar $V_{c,max}$ and J for the two methods (Table 1). However, there were significant offsets in
64 R_d (-95%), Γ (-11%), and $c_{i,trans}$ (-13%) for RACiR measurements (Table 1), consistent with
65 the initial observation of an offset towards unrealistic Γ for C_4 and more negative Γ for C_3
66 leaves measured using RACiR (Fig. 1). Notably, the estimate of R_d from the RACiR data was
67 $0.04 \pm 0.036 \mu\text{mol m}^{-2} \text{s}^{-1}$ (mean \pm s.e.m.), compared with 0.81 ± 0.215 for traditional A/c_i
68 measurements (paired t-test: $t = 3.98$, d.f. = 5, $P = 0.01$).

69 Further model fits were undertaken to determine if estimates of $V_{c,max}$ and J obtained
70 using RACiR were affected by assuming fixed values for all, or only some of the parameters
71 Γ^* , K_{CO} and R_d . If none of the model's parameters were fixed, R_d tended to zero for both
72 RACiR and standard A/c_i measurements (Supplementary Table 1). When we constrained Γ^* ,
73 K_{CO} , and R_d (based on the initial standard A/c_i fit that fixed Γ^* and K_{CO}), $V_{c,max}$ increased for
74 the RACiR, resulting in a significant difference compared with the standard A/c_i data (+6%;
75 Table 1). Differences in $c_{i,trans}$ and Γ also remained significant (Table 1). The difference in
76 $c_{i,trans}$ increased from 13 to 16%, while the difference in Γ decreased dramatically from 11%
77 to almost zero, probably because of the proximity of Γ to Γ^* and R_d .

78 To establish the potential impact of RACiR offsets on Γ^* , we fit an FvCB model in
79 which only K_{CO} and R_d (as above) were held constant. As anticipated, Γ^* from the standard
80 A/c_i responses were close to (<0.01% greater than) values from Bernacchi et al., (2001; Table
81 1). By contrast, Γ^* from RACiR was 20% lower and paralleled the lower Γ (Table 1).
82 Compared with a model that fixed Γ^* , fitting Γ^* while holding R_d and K_{CO} constant resulted
83 in smaller differences in $V_{c,max}$ (2.7% difference) and $c_{i,trans}$ (9% difference), while not
84 inconsiderable in the case of $c_{i,trans}$ ($39 \mu\text{mol mol}^{-1}$), differences between RACiR and
85 standard A/c_i for these two parameters were not statistically significant. However, fitting Γ^*
86 to the data resulted in significantly lower J (6.5%) for the RACiR than the standard A/c_i .

87 To determine whether estimating both Γ^* and R_d affected the match between $V_{c,max}$
88 and J for the two techniques, a final fit was implemented fixing only the value of K_{CO} . This
89 decreased R_d for both measurement techniques and J estimated from RACiR data, resulting

90 in a significant (6%) difference between the standard and RACiR estimates of J . As expected,
91 Γ^* , Γ and $c_{i,trans}$ were also underestimated in RACiR data (Table 1).

92 These analyses show that for the dataset of Stinziano et al., (2017) fixing Γ^* and K_{CO}
93 for A/c_i fits results in reasonably close agreement of V_{cmax} and J for paired RACiR and
94 standard A/c_i measurements. Caution is still necessary, however, because the offsets in
95 RACiR measurements shown in Fig. 1 mean that RACiR tends to estimate V_{cmax} as higher
96 and J as lower than standard A/c_i measurements. Since rates of flow, CO_2 ramping during
97 RACiR measurements, and instrument-specific match parameters could be factors
98 contributing to these offsets, we suggest that work using the RACiR technique should
99 incorporate ground-truthing controls that compare paired RACiR standard A/c_i
100 measurements. This will allow empirical corrections to be applied to RACiR offsets.

101 The standard A/c_i technique does take longer than RACiR. However, it allows
102 modulated fluorescence measurements and estimation of mesophyll conductance and c_c by
103 the variable J method (Harley et al., 1992) that can separate variation in g_m from variation in
104 V_{cmax} . For investigators focussed on understanding the biochemical factors limiting crop
105 photosynthesis and the impacts of rising $[CO_2]$, it is particularly important to realise that
106 RACiR measurements can result in large underestimates of R_d , Γ , and $c_{i,trans}$. These offsets
107 will impact identification of biochemical factors limiting photosynthesis and affect analyses
108 that aim to partition the impacts of biochemical and stomatal limitations on photosynthesis.

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150 Key words: Photosynthesis, Rubisco, A/c_i, rising CO₂, CO₂ compensation point, global
151 change, crop adaptation, phenomics.

152 **Figure Legends**

153 **Figure 1**

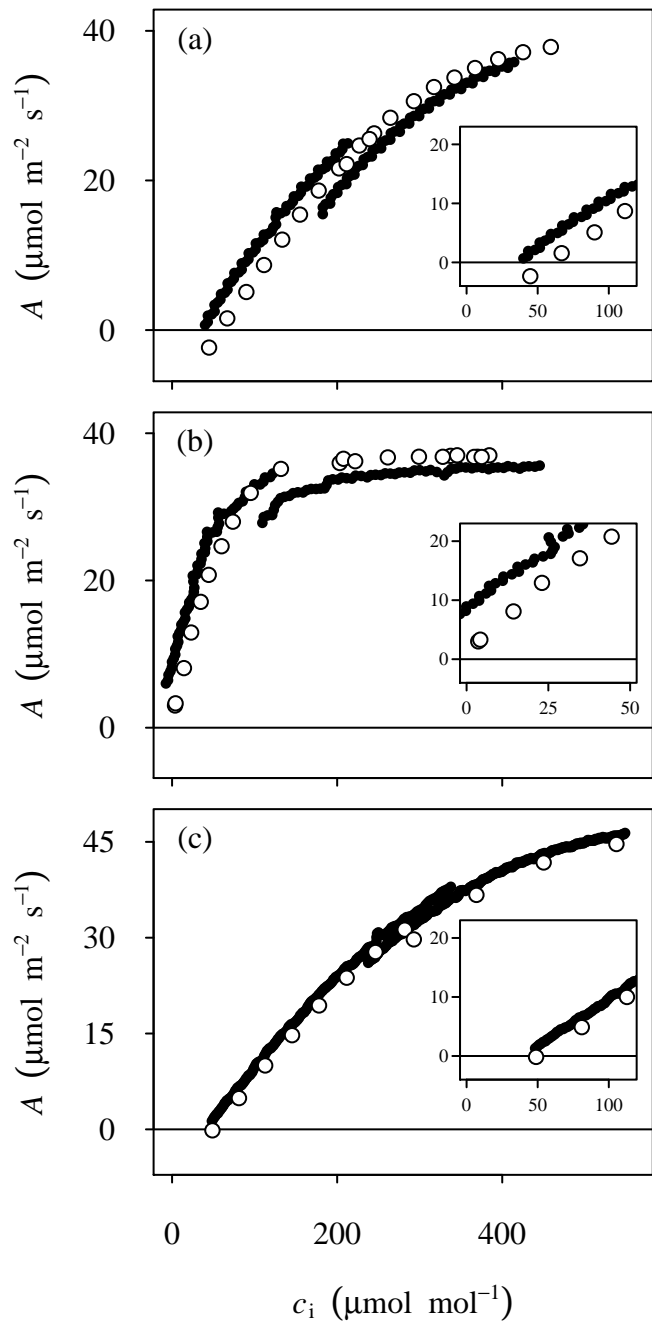
154 Standard A/c_i (open) and RACiR (closed) measurements made for (a) cowpea (*Vigna*
155 *unguiculata*); (b) the C₄ grass teff (*Eragrostis teff*); and (c) selected data for *Populus*
156 *deltoides* from Stinziano et al., (2017). Insets show the region around Γ for each set of
157 measurements. Measurements of *V. unguiculata* and *E. teff* used respective flow rates of 500
158 and 600 $\mu\text{mol s}^{-1}$; incident photosynthetic photon flux density (PPFD), 1500 $\mu\text{mol m}^{-2} \text{s}^{-1}$;
159 air temperature, 30 °C; leaf vapour pressure deficit (VPD_{leaf}), 1.7 kPa; and chamber over-
160 pressure, 0.1 kPa. For *P. deltoides*: flow rate, 600 $\mu\text{mol s}^{-1}$; leaf temperature, 25 °C; VPD_{leaf} ,
161 1.2 kPa; overpressure, 0.2 kPa (Stinziano et al., 2017)

162

163 **Table 1** Parameter choice affects statistical outcomes when comparing standard A/c_i and RACiR measurements made by Stinziano et al., (2017).

Fixed parameters		V_{cmax}	J	R_d	Γ^*	Γ	$c_{i,\text{trans}}$	K_{CO}
Γ^*, K_{CO}	standard A/c_i	110±5.2	230±11	0.81±0.215	42.9	48.6±1.53	432±16.4	714
	RACiR	114±5.1	223±11	0.04±0.036	42.9	43.1±0.265	374±22.6	713
	$\Delta\%$	3.6	-3.0	-95.1	-	-11.3	-13.4	-
	P (paired t-test, d.f.=5)	0.223	0.273	0.011	-	0.011	0.016	-
$\Gamma^*, K_{\text{CO}}, R_d$	standard A/c_i	110±5.3	230±11.3	0.81	42.9	48.58±0.284	434±19.5	714
	RACiR	117±5.3	228±11	0.81	42.9	48.16±0.226	364±22.6	713
	$\Delta\%$	6.4	-0.01	-	-	-0.01	-16.1	-
	P (paired t-test, d.f.=5)	0.014	0.585	-	-	0.029	0.006	-
K_{CO}, R_d	standard A/c_i	110±5.1	231±11	0.81	43.2±1.44	48.9±1.49	431±15.6	714
	RACiR	113±4.6	216±10	0.81	34.2±1.09	39.7±1.25	392±21.1	713
	$\Delta\%$	2.7	-6.5	-	-20.8	-18.8	-9.0	-
	P (paired t-test, d.f.=5)	0.341	0.033	-	4×10⁻⁵	2×10⁻⁵	0.056	-
K_{CO}	standard A/c_i	110±5	232±12	0.34±0.262	46.2±1.68	48.7±1.47	428±16.7	714
	RACiR	112±4.5	218±11	0.3±0.31	37.5±2.75	39.8±1.22	389±20.2	713
	$\Delta\%$	1.8	-6	-11.8	-18.8	-18.3	-9.1	-
	P (paired t-test, d.f.=5)	0.377	0.026	0.934	0.021	3×10⁻⁵	0.049	-

164 Using the R Language and Environment (R Core Team, 2017) function *optim*, the residual sum of squares (RSS) for A_C and A_J was used to
 165 establish best fitting parameter combinations with $c_{i,\text{trans}}$ values from 50 to 499 $\mu\text{mol mol}^{-1}$, then RSS was used to further select the $c_{i,\text{trans}}$ at
 166 which the optimisation had produced the best fit.



Supplementary Material

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Supplementary Table 1 FvCB model fits made allowing all parameters to be estimated from the data, for six paired measurements from the Supplementary data of Stinziano et al., (2017).

ID	Data collection method	V_{cmax}	J	R_d	Γ^*	K_{CO}	Γ	$C_{\text{i,trans}}$
LI1A	Standard	102	262	0	49.1	553	49.1	472
	RACiR	98	240	0	40.6	500	40.6	389
LI1B	Standard	75	180	0.9	41.5	547	48.4	411
	RACiR	91	175	2.7	22.1	588	41.1	350
LI1C	Standard	118	238	0	50.7	786	50.7	425
	RACiR	109	221	0	39.8	709	39.8	411
LI2A	Standard	99	256	0	48.1	500	48.1	410
	RACiR	103	262	0	41.8	500	41.8	427
LI2B	Standard	97	232	0	52.3	598	52.3	417
	RACiR	87	206	0	45.2	500	45.2	340
LI2C	Standard	124	233	1.1	42.5	900	50.9	476
	RACiR	118	218	0	42.1	884	42.1	450