On measuring the response of mesophyll conductance to carbon dioxide with the variable J method

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Abstract

The response of mesophyll conductance to CO2 (gm) to environmental variation is a challenging parameter to measure with current methods. The ‘variable J’ technique, used in the majority of studies of gm, assumes a one-to-one relationship between photosystem II (PSII) fluorescence and photosynthesis under non-photorespiratory conditions. When calibrating this relationship for *Populus trichocarpa*, it was found that calibration relationships produced using variation in light and CO2 were not equivalent, and in all cases the relationships were non-linear—something not accounted for in previous studies. Detailed analyses were performed of whether different calibration procedures affect the observed gm response to CO2. Past linear and assumed calibration methods resulted in systematic biases in the fluorescence estimates of electron transport. A sensitivity analysis on modelled data (where gm was held constant) demonstrated that biases in the estimation of electron transport as small as 2% (~0.5 μmol m−2 s−1) resulted in apparent changes in the relationship of gm to CO2 of similar shape and magnitude to those observed with past calibration techniques. This sensitivity to biases introduced during calibrations leads to results where gm artefactually decreases with CO2, assuming that gm is constant; if gm responds to CO2, then biases associated with past calibration methods would lead to overestimates of the slope of the relationship. Non-linear calibrations were evaluated; these removed the bias present in past calibrations, but the method remained sensitive to measurement errors. Thus measurement errors, calibration non-linearities leading to bias, and the sensitivity of variable J gm, hinders its use under conditions of varying CO2 or light.

Key words: Chlorophyll fluorescence, curve fitting, electron transport rate, gm, mesophyll conductance to CO2, *Populus trichocarpa*, variable J technique.

Introduction

Mesophyll conductance (gm) is the conductance of CO2 from the intercellular airspaces to Rubisco, a largely liquid pathway through the cell wall and three membranes. Whether gm is a constitutive or dynamic characteristic of a leaf is fundamental to our understanding of plant responses to the environment. As gm may represent up to...
40% of the CO₂ diffusional limitation on photosynthesis (Warren, 2008), dynamic variation in gₘ would offer a major avenue for photosynthetic regulation, comparable with that of the stomata. The most commonly used technique to measure gₘ, the variable J method, consistently demonstrates a large reduction in gₘ with increasing CO₂ (Flexas et al., 2007). However, the size and presence of the response of gₘ to CO₂ varies between studies using a variety of methods (Flexas et al., 2007; Tazoe et al., 2009; Vrabi et al., 2009). For example, a less steep response of gₘ to CO₂ was found for Nicotiana tabacum using the independent carbon isotope method relative to the variable J method (Flexas et al., 2007). In a separate experiment, Arabidopsis thaliana and N. tabacum were reported to reduce gₘ by ~85% and 65%, respectively, when CO₂ changed from 200 μmol mol⁻¹ to 1000 μmol mol⁻¹ at 21% O₂ and measured using the variable J method (Flexas et al., 2007). In a second investigation, the carbon isotope method resulted in only a 10% reduction and a 5% increase for the same species, respectively, when measured across the same range of CO₂ mole fractions; measurements at 2% O₂ showed reductions of 26% and 40% (Tazoe et al., 2011). The widely used curve-fitting techniques for estimating gₘ explicitly assume a constant gₘ across the range of CO₂ used to generate CO₂ response curves (Ethier et al., 2006; Warren, 2006; Sharkey et al., 2007; Gu et al., 2010). To date the underlying mechanisms determining gₘ and the reason for the different results between methods remain unresolved.

The ‘variable J’ technique encompasses a group of methods that estimate gₘ, chloroplastic CO₂ concentration (C_c), and the rate of oxygenation or photorespiration (V_o) from combined fluorescence and gas exchange data. Mesophyll conductance to CO₂ is calculated as the ratio of net photosynthetic CO₂ flux (A) to the difference in CO₂ concentration between the intercellular airspaces (C_i) and the chloroplast (C_c). C_c is related to the rate of carboxylation to photorespiration at Rubisco, and photorespiration is then proportional to the difference between fluorescence-derived estimates of the total electron transport rate and the rate of electron use by carboxylation estimated from gas exchange. This derivation is described in detail in the Materials and methods and reviewed by Warren (2006) and by Pons et al. (2009).

Fluorescence estimates of total electron transport are derived from the work of Genty et al. (1989), who established that under non-photorespiratory conditions (low oxygen and high CO₂), a linear relationship exists between the quantum yield of fluorescence (Φ_PSII) and measured quantum efficiency of rates of CO₂ fixation (ΦCO₂). This proportionality has subsequently been used to provide an estimate of electron transport rates: in the absence of alternative electron sinks, the relationship between carboxylation estimates of linear electron flow and fluorescence estimates of electron transport should be one-to-one under non-photorespiratory conditions. In practice, this relationship deviates from one-to-one due to interspecific variation in the values of standard constants such as leaf absorptance (Baker, 2008), and measurement of the relationship under non-saturating CO₂ where significant alternative electron transport sinks may be present. However, for simplicity, it is often assumed that standard constants are accurate and do not vary during experiments.

An alternative approach is to conduct pre-experimental calibrations to provide estimates of electron transport from photosystem II (PSII) fluorescence (Lawlor and Tezara, 2009). While empirical calibration has the potential to improve estimates of electron transport, it can also introduce systematic errors (biases) in the calculation of the total electron transport rate, and thus gₘ. The impact of calibration issues, such as non-linearity, on the calculation of gₘ has not been thoroughly assessed.

The present study examined whether the differences between two common methods used for measuring gₘ is the result of biases in the calibration of the variable J method. However, the challenges inherent in the variable J method have long been recognized (Harley et al., 1992), with the Harley criterion providing an indication of how sensitive the gₘ values are to errors when using this method (Harley et al., 1992). The original sensitivity analyses of Harley et al. (1992) demonstrated that the relationship of gₘ to CO₂ was sensitive to errors in the values of mitochondrial respiration, the photo-compensation point, and the fluorescence estimate of the total electron transport rate. However, this analysis was not extended to a broad range of C_i's, as subsequent studies do, and the sensitivity of the gₘ response to CO₂ to errors has not been compared with the size of biases present in the calibration procedure.

Therefore, the goal of this study was to understand the conditions under which gₘ can be accurately measured using the variable J technique. Fluorescence with gas exchange measurements is calibrated using classical methods for the widely used genome model plant poplar (Populus trichocarpa Torr. & Gray). Consistent with the original literature, significant variation in the calibration relationship was found, such that there is the potential for systematic error when calibrations are applied to a broad range of environmental conditions. Photosynthetic models are then used to assess the effects of biases on the response of gₘ to CO₂. Finally, new calibration techniques by which these biases may be reduced when estimating a single value of gₘ for a leaf, or comparing species, are suggested. However, it is demonstrated that the variable J method should be used with caution when measuring the response of gₘ to CO₂ and light, as any bias in the estimation of electron transport rates results in changes in the relationship of gₘ with CO₂.

Materials and methods

Plant material and growing conditions

Poplar plants were propagated from cuttings and grown in environmentally controlled growth chambers. Metal–halide and high pressure sodium lighting (400 μmol m⁻² s⁻¹) was provided for 14 h per day. Temperatures in the chambers were maintained...
between 20 °C and 24 °C, and humidity was kept at 70%. The cuttings were placed in 3785 cm² pots in Farad 3B potting soil which included Osmocote Plus slow release fertilizer as per the manufacturer’s instructions (15/9/12/1 N/P/K/Mg plus trace elements: S, B, Cu, Fe, Mn, Mo, and Zn; Scotts Company, OH, USA). The pots were watered daily and fertilized weekly with Peters Excel All Purpose soluble fertilizer (21/5/20 N/P/K plus trace elements: B, Cu, Fe, Mn, Mo, and Zn; Scotts Company). Plants were measured after 4-9 months of growth in March–May of 2010 (all experiments), and a second set of plants in November 2010 (extra CO₂ response curves).

Gas exchange and fluorescence measurements

Gas exchange and fluorescence measurements were done on young fully expanded leaves using a 2 cm² LI-COR LI-6400 fluorescence chamber and gas exchange system (LI-COR, Lincoln, NE, USA). Plants were allowed to acclimate to the gas exchange system in a laboratory growth chamber for >30 min, until stomatal conductance was stable. Unless otherwise noted, general measurement conditions were as follows: photosynthetic photon flux density (PPFD), 400 µmol m⁻² s⁻¹ with no blue light component (Loreto et al., 2009); reference CO₂, 400 µmol mol⁻¹; Tₑ, 24.9 ± 0.8 °C; vapour pressure deficit (VPD), 1.47 ± 0.49 kPa; flow, 150 µmol s⁻¹. Single flash fluorescence measurement settings were used and adjusted according to the optimal values obtained from the flash and measuring intensity procedures in the LI-6400 manual (Anon, 2004). All measurements were corrected for leaks using empirically determined leak corrections for dry poplar leaves under measurement conditions (CO₂corrected = CO₂uncorrected + 9.868 × 10⁻² × CO₂R² + 5.29 × 10⁻⁴ × CO₂R² – 1.008).

The Laser method was used to measure non-photorespiratory respiration in the light (Rₐ) and the apparent photo-compensation point (Cₚ) (Warren, 2006). In this method, the y and x value of the average intersection of three CO₂ responses are taken as Rₐ and Cₚ. The CO₂ responses were measured at reference CO₂ concentrations of 150, 100, 75, and 50 µmol mol⁻¹, at three light levels 400, 175, and 75 µmol m⁻² s⁻¹. Six replicate sets of measurements were made providing mean (±SD) values of 0.42 ± 0.21 µmol m⁻² s⁻¹ for Rₐ and 36.0 ± 1.9 µmol mol⁻¹ for Cₚ. In theory, the Cₚ values should be increased by Rₐ/gₘ, to obtain an estimate of the photocompensation point τ (von Caemmerer, 2000), but as no independent value for gₘ was available, the transformation was not performed and gₘ was taken to be equal to Cₚ. A sensitivity analysis, described below, confirmed that minor variation in the value for τ did not greatly affect the value of gₘ (Table 1).

Calibration relationships and CO₂ responses

Light or CO₂ response curves were measured under non-photorespiratory conditions to calibrate the relationship between fluorescence-derived electron transport rates (Jₑₑₑₑ) and photosynthesis (Jₐ₋ₐ). Nine light response curves were measured at ambient CO₂ (400 µmol mol⁻¹) and 1% O₂, starting at a PPFD of 2000 µmol m⁻² s⁻¹ and reducing PPFD to 1500, 1000, 800, 600, 500, 400, 350, 300, 250, 200, 150, and 100 µmol m⁻² s⁻¹ with 4-6 min intervals between measurements. Five CO₂ response curves were measured at 400 µmol m⁻² s⁻¹ PPFD and three at

### Table 1. Sensitivity analysis for the calibration and calculation of gₘ at ambient CO₂ (400 µmol mol⁻¹) and 400 µmol m⁻² s⁻¹ PPFD

<table>
<thead>
<tr>
<th>Calibration type</th>
<th>Parameter varied</th>
<th>Apparent gₘ (mol m⁻² s⁻¹)</th>
<th>% increase with calibration</th>
</tr>
</thead>
<tbody>
<tr>
<td>No calibration standard parameters</td>
<td>1a α = 0.85, β = 0.5</td>
<td>NA</td>
<td>0.166 ± 0.018</td>
</tr>
<tr>
<td>1b α = 0.85, β = 0.5</td>
<td>Rₐ+95% CI</td>
<td>0.173 ± 0.020</td>
<td>4.1</td>
</tr>
<tr>
<td>1c α = 0.85, β = 0.5</td>
<td>Rₐ+95% CI</td>
<td>0.160 ± 0.017</td>
<td>-3.7</td>
</tr>
<tr>
<td>1d α = 0.85, β = 0.5</td>
<td>Γ +95% CI</td>
<td>0.181 ± 0.021</td>
<td>9.0</td>
</tr>
<tr>
<td>1e α = 0.85, β = 0.5</td>
<td>Γ +95% CI</td>
<td>0.154 ± 0.016</td>
<td>-7.5</td>
</tr>
<tr>
<td>Measured leaf absorbance</td>
<td>2 α = 0.831 (measured), β = 0.5</td>
<td>NA</td>
<td>0.182 ± 0.022</td>
</tr>
<tr>
<td>3a Linear fit</td>
<td>Rₐ</td>
<td>NA</td>
<td>0.341 ± 0.087</td>
</tr>
<tr>
<td>3b Linear-sigmoidal</td>
<td>Rₐ</td>
<td>NA</td>
<td>0.306 ± 0.079</td>
</tr>
<tr>
<td>CO₂ response at 1% O₂ and 400 µmol mol⁻¹ s⁻¹ light calibration</td>
<td>4 Linear-sigmoidal</td>
<td>NA</td>
<td>0.199 ± 0.017</td>
</tr>
</tbody>
</table>

Rₐ and Γ were estimated from six replicate sets of Laisk curves: in 1b–1e the mean values were used plus or minus the 95% confidence interval of the mean for the replicates Rₐ, 0.42 ± 0.17 µmol m⁻² s⁻¹ and Γ, 36.0 ± 1.5 µmol mol⁻¹. Linear fit to high light data from the efficiency plot (Fig. 1C) for data where Φₑₑₑₑ < 0.05.

1000 µmol m⁻² s⁻¹ PPFD at 1% O₂, CO₂ was reduced from 400 µmol mol⁻¹ to 100 µmol mol⁻¹ in decrements of 75 µmol mol⁻¹, and after an 8 min re-acclimation at 400 µmol mol⁻¹ increasing CO₂ to 600, 800, 1000, 1500, and 2050 µmol mol⁻¹. Gas exchange was measured at each CO₂ concentration after the cuvette CO₂ concentration was stable for >120 s. A second similar series of CO₂ responses was measured after the first under 21% O₂. The leaf absorbance of 10 leaves was measured using a Taylor integrating sphere (LI-COR 1800-12).

Estimation of ‘variable J’ gₘ

Values for gₘ were estimated from the following standard formulae used in the ‘variable J’ technique (Harley et al., 1992; Valentini et al., 1995; von Caemmerer, 2000), and using variants of the calibrations detailed below. Mesophyll conduc-
tance to CO₂ is estimated as the ratio of the net photosynthetic rate (A) and the difference in CO₂ mole fraction from the intercellular airspaces (Cᵢ) and the chloroplastic sites of photosynthesis (Cᵥ):

\[
gₘ = \frac{A}{Cᵢ - Cᵥ} \tag{1}\]

As A and Cᵥ are provided by standard gas exchange measurements, estimation of Cᵢ remains as the difficult-to-measure unknown in this equation. Cᵢ is estimated assuming that Rubisco...
specificity to O₂ and CO₂ ($S_{CO_2}$) remains constant, and that the ratio of the carboxylation rate ($V_c$) to the oxygenation rate ($V_o$) varies in direct proportion to the concentration of CO₂ at the site of carboxylation ($C_C$) in the chloroplast or the concentration of oxygen (O) which is assumed not to vary. Thus:

$$C_c = \frac{V_c2\Gamma^*}{V_o}$$

(2)

where $\Gamma^*$ is the photo-compensation point (=0.5×O/$S_{CO_2}$), measured using the Laisk method. $V_o$ can be estimated as the sum of the measured $A$, a value for $R_d$ assumed to be constant and equal to that measured using the Laisk method, and half of $V_c$:

$$V_o = A + R_d + 0.5V_c$$

(3)

$V_c$ is included in $V_o$, as for every two oxygenations one CO₂ is released, leading to gross photosynthesis being underestimated by $A$. The total electron transport ($J_{total}$) is the sum of the reductant required for $V_c$, $V_o$, and any alternative electron transport sinks ($V_{alt}$). Under many conditions four electrons are used per CO₂ molecule fixed (Baker, 2008), and it is known that two photorespiratory cycles release one CO₂, thus the rate of photorespiration can be estimated by rearranging this equation:

$$V_o = J_{total}/4 - (V_c + V_{alt}) = \frac{2}{3}(J_{total}/4 - (A + R_d + V_{alt}))$$

(4)

Here $J_{total}$ includes $V_{alt}$, by calibration of the total electron transport rate estimated from fluorescence ($J_{raw}$) with measurements of $A + R_d$ under non-photorespiratory conditions—where $V_{alt}$ and $V_o$ are assumed to be absent—a calibrated electron transport rate ($J_{cal}$) can be obtained. Under photorespiratory conditions $J_{cal}$ then represents the sum of $V_c$ and $V_o$, such that:

$$V_o = J_{cal}/4 - V_c = \frac{2}{3}(J_{cal}/4 - (A + R_d))$$

(5)

From a theoretical perspective $C_c$ is relatively well defined, but see Parkhurst (1994) and Evans (2009) for issues with describing CO₂ fluxes or fluorescence with an average number representing different depths in the leaf. However, it is the practical estimation of $J_{cal}$ and $V_{alt}$ that remains controversial and which represents a potential source of error in the calculation of $g_m$. To obtain an accurate value for $J_{cal}$, the raw measurements of chlorophyll fluorescence ($J_{raw}$) must be calibrated and in doing so account for $V_{alt}$ under the experimental conditions as follows. Fluorescence of PSII provides an initial estimate of total electron flux through the electron transport chain:

$$J_{raw} = 0.425PPFD\Phi_{PSII}$$

(6)

where 0.425 is the product of 0.85, the standard assumed value for leaf absorbance ($\alpha$), and 0.5, the standard fraction of quanta absorbed by PSII relative to PSI ($\beta$), and $\Phi_{PSII}$ the quantum efficiency of PSII measured from fluorescence [($\Phi_{PSII}=\{(F_{m}-F_{i})/F_{m}\}$] If measured values for leaf absorbance are available, the assumed $\alpha$, and the estimate for $J_{raw}$, can be improved. However this procedure described below are often used to estimate a value for $\alpha\beta$ and therefore $\alpha$ is not typically necessary. $J_{raw}$ then can be related to $J_{A + R_d}$ under appropriate non-photorespiratory conditions—normally at 1% O₂—where $V_o$ is negligible. From this relationship, the empirical values for $\alpha\beta$ can be found and thus provide a calibrated estimate of total electron flux ($J_{cal}$). Under non-photorespiratory conditions Equations 5 and 6 become:

$$J_{cal} = 4(A + R_d) = mL_{raw} + c = m0.425PPFD\Phi_{PSII} + c$$

(7)

assuming a linear relationship. Thus the corrected value for $\alpha\beta$ is $m \times 0.425$. Alternatively, this equation is often converted from voltage to quantum efficiencies by solving for $\Phi_{PSII}$, preferably when no intercept is present:

$$\Phi_{CO_2} = \frac{A + R_d}{PPFD} = \frac{m0.425}{4 - \Phi_{PSII}} = m\Phi_{PSII}$$

(8)

where $\Phi_{CO_2}$ is the quantum efficiency of photosynthesis ($A/R_d$)/PPFD), $m$ is the slope of the efficiency relationship, and the calibrated value for $\alpha\beta$ is $4m'$. In practice, either of these relationships (Equation 7 or 8) are used for the calibration of $J_{raw}$, with the fitted slopes providing an estimate of the value of $\alpha\beta$ for the calibration conditions. The intercept is usually assumed to be zero. Alternatively, the presence of a non-zero $\gamma$ intercept can be tested: if present, it represents alternative electron transport at the photo-compensation point.

This calibration procedure is based upon the assumptions that: (i) $\alpha$ and $\beta$ are constant across the range of experimental variation; (ii) it is possible to estimate alternative electron transport as a constant proportion of total electron flux estimated as the intercept of the relationship; and (iii) the non-photorespiratory measurement conditions do not alter alternative electron transport relative to the experimental conditions. If either $\alpha\beta$ or alternative electron transport vary with the environmental condition used to create the relationship (light or CO₂), non-linearities should be present in the relationship. An alternative is then to fit a non-linear function to the calibration data, such as the following linear–sigmoidal function:

$$J_{cal} = J_{A + R_d} = J_{raw} - c - a/[1 + exp(-(J_{raw} - b)/d)]$$

(9)

Analysis of sensitivity of ‘variable J’ $g_m$, magnitude to calibration scenarios

To test whether calibration variants have significant effects on the calculation of $g_m$, a sensitivity analysis was performed. Apparent shifts in $g_m$ due to changing the calibrations were calculated as follows for gas exchange measurements made on six leaves under ambient conditions (400 μmol mol⁻¹ CO₂ and a PPFD of 400 μmol m⁻² s⁻¹). (1) Standard calibration using assumed $\alpha$ and $\beta$ values (0.85 and 0.5) as is often used for the variable J method, with the following variants: (1a) the mean $R_d$ and $\Gamma^*$ values measured using the Laisk method with six replicates; (1b) $R_d$ plus and (1c) $R_d$ minus the 95% CI of the mean; (1d) $\Gamma^*$ plus and (1e) $\Gamma^*$ minus the 95% confidence interval of the mean. (2) Standard calibration using a measured $\alpha$ (0.831) and assumed $\beta$ value (0.5). (3) Calibrations fit to light response data measured under non-photorespiratory conditions at ambient CO₂ (3a) using a linear fit, passing through the origin on the efficiency plot, but only using data points below a $\Phi_{CO_2}$ of 0.05 as suggested by Seaton and Walker (1990) and (3b) a linear–sigmoidal fit to the combined light response data on the rate plot. (4) Calibrations fit to the CO₂ response data measured at 400 μmol m⁻² s⁻¹ PPFD and under non-photorespiratory conditions, using the linear–sigmoidal function. Fitted parameters for the calibration functions are provided in the Results. The non-linearity of the calibrations was assessed by comparing the Akaake Information Criterion (AIC) values between linear–sigmoidal fits and linear fits, where fits with the lowest AIC values have greatest support with model complexity taken into account (Burnham and Anderson, 2004). The R statistical program was used for these analyses (R_Development_Core_Team, 2010). An apparent value for $g_m$ was calculated for each of the six replicate leaves for all of the scenarios or parameter changes described above.

Cross-validation of ‘variable J’ $g_m$, with $g_m$ estimated from curve-fitting procedures

Values of mesophyll conductance to CO₂ were measured for an additional 10 CO₂ response curves using the same apparatus, corrections, and measurement conditions as detailed above. Added
to the five initial CO₂ response curves measured at 21% O₂, these provided a total of 15 curves, with an average of 14 CO₂ levels per curve. The measurements for the CO₂ response curves were made simultaneously with the fluorescence measurements, by using the 2 cm² LI-COR fluorescence chamber, a necessary compromise, as the goal of this experiment was cross-validation between the variable J and curve-fitting methods. Ideally, measurements for use in curve fitting should be made using larger leaf areas (Warren, 2006; Pons et al., 2009).

The Exhaustive Dynamic Optimization (EDO) curve-fitting technique of Gu et al. (2010), as implemented on the LeafWeb website, was employed for this analysis in cognizance of the curve-fitting parameterization issues raised in that paper. The technique is based upon the principle that the photosynthetic CO₂ response curve can be represented by the minimum of a combination of three equations (Equation 10). These equations are non-rectangular hyperbolas that explicitly account for a non-infinite gₘ, and are based upon the original Farquhar–von Caemmerer–Berry-type photosynthetic functions:

\[ A = \min(A_{\text{a}}, A_{\text{j}}, A_{\text{PPF}}) \] (10)

The EDO approach uses functions for these three processes approximately similar to past curve-fitting approaches (Ethier et al., 2006; Warren, 2006), but applies these by assessing the possibility that any CO₂ response curve point could be limited by any of the three processes, with some constraints. The technique then exhaustively searches for parameter estimates for all of these possible limitation states, and selects the optimal fit as the fit with a minimum of a cost function consistent with the form of Equation 10. For the EDO analysis, five parameters (\(g_{\text{m}}, V_{\text{max}, J}, R_{\text{d}}\) and the rate of triose phosphate utilization) were fit. To test for reliability of the parameter estimate for \(g_{\text{m}}\) the first and second derivatives of the cost function with respect to \(g_{\text{m}}\) were tested to be zero and non-zero, respectively (Gu et al., 2010). To enable comparison between the two methods, \(g_{\text{m}}\) values for the curve-fitting procedure (representing the entire CO₂ response curve) were compared with variable \(g_{\text{m}}\) values measured at ambient CO₂ (ambient CO₂ point on the CO₂ response curve), or the interpolated \(g_{\text{m}}\) value for a \(C_{i}\) of 600 μmol mol⁻¹ (interpolated as the point at a \(C_{i}\) of 600 μmol mol⁻¹ on the line connecting the measured \(g_{\text{m}}\) and \(C_{i}\) value greater and less than 600 μmol mol⁻¹). Sensitivity analysis of ‘variable J’ \(g_{\text{m}}\) to CO₂ variation

The sensitivity of \(g_{\text{m}}\) response to variation in CO₂ in the estimation of \(R_{\text{d}}\) or \(J_{\text{cal}}\) was assessed by introducing a constant offset into a Farquhar–von Caemmerer–Berry-type photosynthetic model that held \(g_{\text{m}}\) constant. A region of the modelled \(g_{\text{m}}\) to \(C_{i}\) response curve was defined as a \(C_{i}\) of 200–500 μmol mol⁻¹ corresponding to values for which photorespiration should not be greatly inhibited, and thus measurements of Uₐ would be relatively accurate. A linear slope was fit to these data, and non-zero slopes were recorded in response to introducing positive or negative biases in \(R_{\text{d}}\) or \(J_{\text{cal}}\).

The photosynthetic modelling was conducted using inputs of varying \(C_{i}\) and constant values of \(g_{\text{m}}\) (0.3 mol m⁻² s⁻¹), \(R_{\text{d}}\) (0.42 μmol m⁻² s⁻¹), \(\Gamma_{\text{s}}\) (36.0 μmol mol⁻¹), \(V_{\text{max}}\) (70 μmol m⁻² s⁻¹), and \(J\) (108 μmol m⁻² s⁻¹). The values for these parameters were chosen to represent approximately a measured CO₂ response curve for P. trichocarpa. The model calculations are provided online as a spreadsheet (Supplementary Spreadsheet S1 available at JXB online). \(K_{\text{s}}\) and \(K_{\text{a}}\) values and the standard Farquhar–von Caemmerer–Berry equations were taken from von Caemmerer (2000), and reference to them is provided in the spreadsheet. From these inputs, \(V_{\text{min}}\) was calculated from the limiting process, namely the minimum of \(V_{\text{c,s}}\) and \(V_{\text{c,f}}\) (the Rubisco and RuBP regeneration-limited carboxylation rates), and the photorespiration rate, \(V_{\text{r}}\), calculated from \(V_{\text{min}}\). The total RuBP regeneration rate, \(J_{\text{cal/4}}\), was calculated as the sum of \(V_{\text{c,min}}\) and \(V_{\text{r}}\). Sensitivity analysis of ‘variable J’ \(g_{\text{m}}\) to CO₂ variation

Results

Calibration of fluorescence estimates of electron transport rate with gas exchange

Data used to calibrate fluorescence with gas exchange can be expressed either as quantum efficiency plots or as rate plots using CO₂ or electron equivalent units. As each has its advantages, the same light or CO₂ response curve data were compared on both plots (Fig. 1). The calibration relationships relating photosynthetic rates in electron-equivalent units (\(J_{\text{A+R}}\)) to uncalibrated fluorescence estimates of electron transport (\(J_{\text{raw}}\)) under non-photorespiratory conditions were non-linear when measured across a broad range of light or CO₂ conditions (Fig. 1). Light response curves demonstrated three phases of non-linearity: a subtle increase in \(J_{\text{raw}}\) relative to \(J_{\text{A+R}}\) at low light, a large shift towards increased \(J_{\text{raw}}\) but not \(J_{\text{A+R}}\) at intermediate light, and in some responses a return to the one-to-one line at the highest light levels (Fig. 1A). For the same data plotted as efficiency plots (note the reverse in direction representing increasing light), the same shifts resulted in curvature towards greater quantum efficiency of net photosynthesis (\(\Phi_{\text{C0}}\)) at low light (Fig. 1C). This method of plotting the same data emphasizes the second curvature towards greater PSI efficiency (\(\Phi_{\text{PSI}}\)) at very low light (~100 μmol m⁻² s⁻¹).

Due to the curvature of the light response data from low to high light, two calibration relationships were fit. In the first method, a linear calibration was fit to each replicate light response on the efficiency plot forcing each line to pass through the origin (intercepts were not significantly different from the origin over this range of PPFD) and using data below a 0.451 rather than 0.5. The measured value of \(\alpha\) was 0.831, resulting in an estimate of 0.461 for the \(\beta\) value for higher light intensities. For comparative purposes, the one-to-one line was considered as the standard ‘calibration’ (\(\alpha=0.425\), as it is common to assume this value for \(\alpha\) with no further calibration. In the second calibration, a linear–sigmoidal curve was fit to all of the light response replicates simultaneously for the rate plot (Equation 9, a linear–sigmoidal fit: \(a=11.1, b=99.9, c=1.87, d=8.31, \text{ adjusted } R^2=0.974, \text{ AIC value}=597.47\). A linear fit
to the same data resulted in a marginally higher AIC value (598.81) and similar adjusted R² (0.974).

It is common to use light response curves to calibrate the variable J method and then make use of them in studies using other experimental stimuli, for example variation in CO₂. The carbon dioxide response curves measured at low O₂ and 400 μmol m⁻² s⁻¹ PPFD did not resemble the light response curves at lower CO₂ concentrations, or at higher light (Fig. 1B). A large shift was observed in the opposite direction to the light response curves, consistent with an increase in alternative electron sinks as may be expected by the lack of photosynthetic ability to use reductant under conditions of high light, low CO₂, and low O₂. As the high CO₂ points were measured after the low CO₂ points, and showed high efficiency nearing the one-to-one line, photo-inhibition was not apparent (also F'/Fm’ returned to pre-low CO₂ exposure levels). The relationship on the efficiency plot was not linear; therefore, the CO₂ response curves were only calibrated using a linear–sigmoidal function on the rate plots (Equation 9, a linear–sigmoidal fit: a=228.8, b=953.3, c=−0.51, d=−8.6, adjusted R²=0.958, AIC value=357.7). A linear fit to the same data resulted in a considerably higher AIC value (367.6) and lower adjusted R² (0.949), the difference between AIC values of −10 signifying that the
linear–sigmoidal fit had more support than the linear fit, despite taking into account the extra parameters in the linear–sigmoidal model (Burnham and Anderson, 2004). At a higher PPFD of 1000 μmol m⁻² s⁻¹, CO₂ responses showed greater deviation from the one-to-one line, with ~40 μmol e⁻ m⁻² s⁻¹ of apparent electron transport for little assimilation at the lowest CO₂ levels. To the authors’ knowledge, robust data have not been presented to validate that the calibration relationship is the same between conditions of varying light and CO₂, apart from Hassiotou et al. (2009) whose results largely confirm those in Fig. 1. Most studies use a single saturating flash to measure \( F_m' \), potentially introducing additional non-linear effects with changing light (Markgraf and Berry, 1990; Earl and Ennahli, 2004). However, measurements of varying light demonstrated that using multiple saturating flashes rather than a single flash did not linearize the calibration functions, and rather the size of the discrepancy between \( J_{A+R_d} \) and \( J_{raw} \) was slightly enhanced at high PPFDs (\( F_m' \) increases while \( F_s \) remains constant). As the key CO₂ calibrations were done at moderate PPFD (400 μmol m⁻² s⁻¹), these were not affected.

Size of calibration biases on \( J_{cal} \) and \( g_m \)

From the data shown in Fig. 1, five types of calibrations were performed: the three linear or linear–sigmoidal functions, the standard assumed calibration parameters from the literature, and the measured leaf absorbtance and assumed \( \beta \). The magnitude of the errors in \( J_{cal} \) on the rate plots, and particularly the efficiency plots, is both difficult to visualize and hard to relate to the magnitude of the measured quantities. Therefore, data were expressed as the residuals for the rate relationship, rescaled to units of CO₂ uptake, and plotted against the PPFD or CO₂ used to generate the points (Fig. 2A, B). The residuals were calculated as \( J_{cal}/4–A \), where \( J_{cal} \) is the electron transport rate calibrated using one of the five types of calibration.

The linear–sigmoidal calibrations applied to the same environmental variation to which they were fit produced the smallest residuals, and did not have any systematic errors across a broad range of light or CO₂, possibly apart from 2000 μmol m⁻² s⁻¹ PPFD (Fig. 2A). The linear higher light calibration produced few residuals at high light, but consistently underestimated \( A \) by ~1 μmol CO₂ m⁻² s⁻¹ at low light. The standard calibration, assuming \( \alpha \beta=0.425 \), performed poorly, with significant overestimates of \( A \) of up to 4 μmol CO₂ m⁻² s⁻¹ under all but low light conditions. The standard calibration was even worse for low CO₂ conditions (Fig. 2B), resulting in residuals as high as 6 μmol CO₂ m⁻² s⁻¹. The calibration with assumed \( \gamma \) and measured \( \alpha \) had a similar pattern to the standard calibration although the residuals were improved.

The variation between these calibration curves resulted in large differences in apparent \( g_m \) values when applied to measured photosynthetic data for ambient CO₂ and moderate light (Table 1). Values ranged by 104% from a minimum of the standard assumed calibration to that of the linear calibration. Linear–sigmoidal fits to light or CO₂ response curves were intermediate. This variation in apparent \( g_m \) due to the underlying calibration was larger than variation in \( g_m \) caused by changes in \( R_d \) or \( \Gamma^* \) when adjusted by the 95% CIs of the mean values (Table 1).

Correspondence between ‘variable J’ \( g_m \) and curve-fitting \( g_m \)

The \( g_m \) values calculated by the variable \( J \) and EDO approach curve-fitting method were most highly correlated,
and the points nearest the one-to-one line for the CO$_2$ linear–sigmoidal calibration (Fig. 3). The assumed and linear-light calibrations resulted in correlations between the variable $J_m$ and the curve-fitting $g_m$ values, but resulted in greater deviation from the one-to-one relationship. In addition, the linear-light calibration resulted in a negative value of $g_m$, which was removed from the analysis. Two $g_m$ values were removed from all analyses due to the EDO curve-fitting analysis providing high $g_m$ values (>0.5 mol m$^{-2}$ s$^{-1}$; $T_l$, 24.9±0.8 °C; VPD, 1.47±0.49 kPa).

**Response of $g_m$ to CO$_2$**

The response of $g_m$ to CO$_2$ (detrended for stomatal conductance changes by using $C_i$) was highly variable when the five calibration protocols were compared (Fig. 4). In one of the five replicates (Fig. 4A), all five calibrations produced values for $g_m$ in the range of past reports (Ninemets et al., 2009); in the other four replicates the linear and linear–sigmoidal fit to the light response calibration resulted in negative or large (>1) values for $g_m$ at CO$_2$ levels higher than ambient (one representative replicate is shown in Fig. 4B).

As only the standard calibration constants and the linear–sigmoidal fit to the CO$_2$ response calibration gave reasonable values for $g_m$ for all replicates, these two calibration protocols were investigated in greater detail. Using either calibration, $g_m$ showed strong shifts, increasing from the lowest $C_i$ values, remaining stable or slowly decreasing at ambient CO$_2$ values, and decreasing strongly at high $C_i$ (Fig. 5A, B). However, the Harley et al. (1992) criterion was violated for almost all points at high $C_i$. Nevertheless, the points that satisfy the Harley criterion (Harley et al., 1992) demonstrate a consistent negative response of $g_m$ to $C_i$.
The potential for calibration biases to affect the calculation of $g_m$ can be illustrated by plotting $C_i$ versus the parameters used to estimate $C_c$ and $g_m$ (Fig. 6). As $C_i$ increases, $V_o$ tends towards zero due to competitive inhibition of photorespiration by CO$_2$. As a result, $C_c$—proportional to the $V_c/V_o$ ratio—is increasingly vulnerable to biases at high $C_i$. Specifically, as $V_o$ decreases at higher CO$_2$s, any errors in its estimation lead to an inflated $C_c$ (calculated from the $V_c/V_o$ ratio), as $C_c$ tends towards $C_i$, $g_m$ values [calculated from $A/(C_i-C_c)$] rapidly become large. Once $C_c$, estimated with slight errors, is the same or larger than $C_i$, $g_m$ becomes infinite or negative. This explains the variability, high and negative values of $g_m$ in Fig. 4. Furthermore, as $V_o$ tends towards zero, biases in $J_{cal}/4$ due to alternative electron transport sinks (or changes in $R_d$, $\alpha$, or $\beta$) become increasingly important. In other words, small errors in the estimation of $J_{cal}$ have increasing impact on the estimation of $g_m$ at high $C_i$, as $V_o$ becomes small and the error to the $V_o$ ratio increases.

**Sensitivity of response of $g_m$ to CO$_2$**

The sensitivity of $g_m$ to calibration biases was further investigated by introducing small biases in $J_{cal}/4$ into a photosynthetic model. The lines in Fig. 7A represent the apparent $g_m$ response for simulated data for which $g_m$ was held constant, but for which small systematic errors were introduced into the value for $J_{cal}/4$ and $g_m$ then calculated from the biased data. Any overestimation of $J_{cal}/4$ results in a lower $g_m$ and an apparent negative relationship with increasing $C_i$ (Fig. 7A). In contrast, underestimating $J_{cal}/4$ results in a larger apparent $g_m$. The presence of a positive or negative relationship between $g_m$ and $C_i$ was a function of the small constant biases added to $J_{cal}/4$ (Fig. 7B). If, in the photosynthetic model, $g_m$ is assumed to be constant with CO$_2$, then the residuals in Fig. 7B demonstrate that previously used calibration relationships would consistently result in apparent negative $g_m$ responses to CO$_2$, while the linear–sigmoidal CO$_2$ response calibration would result in both negative and positive relationships. If this assumption is true, the sensitivity analysis demonstrates that the bias in $J_{cal}/4$ necessary to result in an artefactual effect of $C_i$ on $g_m$ is small (<0.5 $\mu$mol CO$_2$ m$^{-2}$ s$^{-1}$) relative to the residuals typically observed in the calibration relationship (~0.5–6 $\mu$mol CO$_2$ m$^{-2}$ s$^{-1}$; Fig. 2A or B and Fig. 7B). If, however, $g_m$ is truly not constant, the observed slope of the response of $g_m$ to CO$_2$ would still be sensitive to errors, and
photosynthetic electron transport rate ($J_{cal}$). This was demonstrated using a sensitivity analysis of standard photosynthetic equations, to which a systematic bias was added. For example, the sensitivity is such that there is an apparent 23% decrease in $g_m$ over a 300 μmol mol$^{-1}$ range of $C_i$ when an ~2% (~0.5 μmol CO$_2$ m$^{-2}$ s$^{-1}$) overestimate of $J_{cal}/4$ is included in the photosynthetic model, despite the modelled $g_m$ remaining constant (Fig. 7A). As the true modelled relationship was on the steepest portion of the sensitivity analysis (Fig. 7B), this demonstrates that if $g_m$ is indeed constant, then any bias in $J_{cal}/4$ will lead to artefactual positive or negative relationships of $g_m$ to $C_i$. If $g_m$ is dynamic, varying with CO$_2$, the point of greatest sensitivity will drift, but the overall pattern of sensitivity demonstrated will remain. In this case, an observed relationship may represent a true response, but the slope will be sensitive to measurement errors and calibration biases. This sensitivity analysis provides similar results to those which Harley et al. (1992) presented in their fig. 6, and those which Hassiotou et al. (2009) presented in their supplementary material. Indeed, Harley et al. (1992) note that: ‘In all cases, the sensitivity to errors was relatively low between 100 and 300 μbar $C_i$, but outside this range the sensitivity was so great that the results could become unreliable.’ Despite these earlier cautions, subsequent researchers have continued to use this approach over a broad range of conditions. It is important to note that these considerations are applicable to any environmental variation that may affect photorespiration: CO$_2$, temperature, light, stomatal closure, etc. For instance, a similar analysis could be done for the relationship of $g_m$ to PPFD, in which case the relationship would be sensitive to errors at PPFDs below light saturation where the errors become significant relative to photorespiration. Thus it is also the relationship of $g_m$ to light that is sensitive to errors when using the variable $J$ method, although at saturating light intensities the presence of high rates of photorespiration leads to less sensitive estimates of the relationship of $g_m$ to light.

The residual variation in the different calibration relationships—which is a determinant of the error in $J_{cal}/4$—was up to 5 μmol CO$_2$ m$^{-2}$ s$^{-1}$ at the extreme of using standard calibration constants, and about ±1 μmol CO$_2$ m$^{-2}$ s$^{-1}$ when calibrated using a linear–sigmoidal function on CO$_2$ response data (Fig. 2A, B). Thus, the magnitude of the errors in the calibrations was similar to, or considerably larger than, the error necessary to affect whether there is an apparent response of $g_m$ to $C_i$ (Fig. 7B). It is broadly true then, given the large errors in our best estimates of $J_{cal}/4$, and the sensitivity of the $g_m$ to $C_i$ relationship to any error, that it is difficult to measure the response of $g_m$ to $C_i$ using the variable $J$ method. That is, with the high overestimates of $J_{cal}$ demonstrated for standard calibration methods over a moderate CO$_2$ range (Fig. 7B), the variable $J$ method is likely to produce steeper relationships between $g_m$ and CO$_2$ than actually exist.

Variable $J$ $g_m$ and partially independent $g_m$ values from the EDO curve-fitting approach corresponded well when the variable $J$ technique was limited to use under ambient

if errors were large enough would result in transitions from negative to positive or vice versa. The model is included as a spreadsheet in the Supplementary data at JXB online.

Discussion

Can the variable J method measure the response of $g_m$ to CO$_2$?

The nature of the observed response of $g_m$ to CO$_2$ is highly sensitive to biases in the estimation of the calibrated total

![Figure 7](https://academic.oup.com/jxb/article-abstract/63/1/413/559130/fig7)
CO₂ and with the non-linear calibrations reported here (Fig. 3). These results are consistent with the sensitivity analysis performed earlier (Table 1). That is, relative to the EDO curve-fitting $g_m$ values the linear-light calibration causes overestimates in variable $J$ $g_m$, the CO₂ linear–sigmoidal calibration results in approximate correspondence, and the assumed calibration results in underestimates. This suggests that when appropriately calibrated the variable $J$ method has value for studies comparing species, using unstressed plants at moderate light and ambient CO₂, but should not be applied across a range of environmental conditions under which photorespiration is likely to vary. However, these non-linear calibrations remain empirical and do not address the implication that a non-linear response represents an unaccounted for fundamental change in photosynthetic functioning.

Why are the calibrations non-linear?

It is difficult to provide a retrospective review of whether the non-linearities in the calibration relationships observed here are present in the $g_m$ literature. For example, a literature review of 56 experimental studies of $g_m$, published since 1992, found that 66% of these use the variable $J$ method, and 44% use it as a sole technique. Of these, few studies provided calibration data, and if this was done even fewer calibrated the variable $J$ method using environmental variation appropriate for the experiment at hand. Fewer performed brief sensitivity analyses, and finally no study attempted to calibrate the technique using non-linear functions. However, many of the non-linear effects described here have been previously described by Seaton and Walker (1990) and Oquist and Chow (1992). There are also indications of non-linearities in the calibration relationships used to calculate $g_m$ or $C_c$ (Warren, 2006; Galle et al., 2009; Hassiotou et al., 2009; Loreto et al., 2009).

Seaton and Walker (1990) and Oquist and Chow (1992) demonstrated large non-linearities in light response curves, plotted as efficiency plots, measured under non-photorespiratory, saturating CO₂ conditions and with oxygen electrodes. These curved relationships on efficiency plots are consistent with the sigmoidal patterns found on the rate plots, but there are clear differences in weighting of points between the plots. The reasons for the non-linearities are discussed by Oquist and Chow (1992) and include: (i) changing connectivity of PSII units, leading to more cycling of electrons between chlorophylls; (ii) at low light, mitochondrial respiration ($R_d$) may increase, but in the variable $J$ calculations $R_d$ is assumed to be constant and a single value usually estimated for all conditions from Lausk response curves; (iii) fluorescence parameters may be estimated from slightly shallower populations of chloroplasts than those that fix CO₂, and the contributions of these populations of chloroplasts would change with light intensity (Warren, 2006; Evans, 2009). On the rate plots, possible alternative electron sinks are highlighted, resulting in non-linear shifts in the calibration relationship, and may represent little (Ruukska et al., 2000), or up to 24% of the total electron flux (Haupt-Herting and Fock, 2002). Two main processes are thought to account for alternative electron sinks (von Caemmerer, 2000), each accounting for up to 10% of total electron flux: reductant provided to nitrate assimilation (Rachmilevitch et al., 2004) and the Mehler reaction (Haupt-Herting and Fock, 2002).

These effects are highlighted when comparing light and CO₂ responses measured under non-photorespiratory conditions (Fig. 1A, B). A priori this must be expected, as at low CO₂, and particularly at high light, there is a limitation on reductant use, but high reductant supply that will result in large alterations of PSI heat dissipation and may result in up-regulation of alternative dissipative energy sinks, such as the Mehler reaction (Neubauer and Yamamoto, 1992). The quantitative effects of alternative energy sinks remain unclear (Ruukska et al., 2000); however, it is noted that relative to the errors (~2% of $J_{total}/4$) necessary to cause apparent changes in $g_m$, estimates of alternative electron sinks are large and therefore vital to account for.

Finally, it is not clear whether alternative electron sinks are changed when shifting from ambient to low O₂ as required for the calibration curves (Pons et al., 2009). For instance, at high CO₂ the calibration curve was closer to the one-to-one line than for high light points (Fig. 1). This may imply that alternative electron transport sinks are affected by the capacity of photosynthesis to dissipate absorbed light energy, or are directly affected by CO₂. Considerable shifts in nitrate assimilation with age, CO₂, and oxygen concentration occur in Arabidopsis, using an equivalent electron flux up to 10% of the photosynthetic rate (Rachmilevitch et al., 2004), and thus represent evidence of alternative electron transport shifts that could occur during the calibration procedure. If this is generally the case, it would be challenging to find conditions under which the variable $J$ method can be calibrated. Indeed, the fitted $\alpha$ or $\beta$ parameters for a non-linear calibration function cannot then be interpreted as physical constants as the non-linearity implies that they change with environmental conditions, or that they include alternative electron transport sinks. It appears that much work remains to be done, using independent methods, to understand the implications of the photosynthetic changes that occur when producing calibration relationships for the estimation of $g_m$ and $C_c$ using the variable $J$ method.

How should the variable $J$ method be used?

The variable $J$ method appears difficult to validate under circumstances of varying photorespiration due to the extreme sensitivity of $g_m$ under conditions of low photorespiration. However, the method when calibrated taking non-linearities into account did improve the estimates of $g_m$ under ambient CO₂ relative to the EDO curve-fitting approach. Thus if the variable $J$ method is to be used for comparing species (and not environmental variation) the following are imperative: (i) a calibration is done with conditions that match the experimental conditions (not a light calibration versus CO₂ experiment); (ii) the
calibration (and experiment) is limited to the linear region, for example $\Phi_{CO_2}<0.05$; Seaton and Walker (1990), or nonlinear functions are used, and if not linearity should be explicitly tested; (iii) the calibration is fit using rate plots, not the efficiency plots that emphasize low photosynthetic rate points disproportionately; and (iv) a sensitivity analysis is done that asks what size biases in the estimation of $J_{cal}$, or variation in the values for $R_d$ and $\Gamma^*$, are necessary to remove the observed effect or relationship, and are such errors plausible for the calibrations. Regardless of these improvements, the lack of knowledge of why the calibration response is curved, and whether alternative electron sinks are affected by changing $O_2$ may preclude the use of the variable $J$ method in most experiments.

**Conclusion**

The variable $J$ method is sensitive to errors and must be used with caution in experiments where photorespiration varies. Nevertheless, none of the calibration or sensitivity scenarios tested here precludes an effect of any variable on $g_m$; thus $g_m$ may be dynamic rather than constitutive, but these results suggest that we cannot know the magnitude or nature of changes with certainty using this technique. It is suggested to limit use of the variable $J$ method to comparing species under conditions of moderate light and ambient $CO_2$ with appropriate calibration, and not in experiments measuring responses to environmental factors that affect photorespiration. There is much research needed using independent methods to provide information on whether and how $g_m$ and alternative electron sinks respond to $CO_2$, light, or $O_2$. The region at which $g_m$ measured using the variable $J$ method starts declining with $CO_2$ (Flexas et al., 2007) or reduced light (personal observation) corresponds to the point where RuBP regeneration becomes limiting to photosynthesis. Although this may occur through a common mechanism related to RuBP regeneration, this effect is less apparent in $g_m$ measurements using carbon isotope discrimination (Flexas et al., 2007; Tazoe et al., 2009, 2011; Vrabl et al., 2009). The point where RuBP regeneration becomes limiting for both the light and $CO_2$ response curves also corresponds to a decrease in photorespiration. Thus at this point the ratio of biases to photorespiration dramatically increases, causing artefacts to be introduced into the response of variable $J$ $g_m$ to $CO_2$ or light, if subtle biases are present in the calibration or measurements. Thus, it is suggested that positive biases in the calibration procedure result in the variable $J$ method overestimating the slope of the relationship between $g_m$ and $C_i$—an explanation for the differences between studies using the variable $J$ method and those using carbon dioxide discrimination.

**Supplementary data**

Supplementary data are available at *JXB* online.

**Spreadsheet S1.** Sensitivity analysis of modelled photosynthetic response to $CO_2$, with $g_m$ held constant.

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**References**


