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Title: The light response of mesophyll conductance is controlled by structure across leaf profiles

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SUMMARY STATEMENT

Using theoretical and observed evidence of the response of mesophyll conductance ($g_m$) to light, it is shown that this response is apparent, where the bulk leaf $g_m$ appears to respond to light while layer-specific $g_m$ values do not. This was successfully represented using a multi-layer leaf model coupled with anatomical observations. This apparent response has implications for how limitation analyses are conducted and illustrates the importance of measuring $g_m$ under saturating light. Mesophyll conductance is an emergent property of the 3D leaf structure and not solely a leaf area based phenomenon.
ABSTRACT

Mesophyll conductance to CO₂ ($g_m$) may respond to light either through regulated dynamic mechanisms or due to anatomical and structural factors. At low light, some layers of cells in the leaf cross-section approach photocompensation and contribute minimally to bulk leaf photosynthesis and little to whole leaf $g_m$ ($g_{m,leaf}$). Thus, the bulk $g_{m,leaf}$ will appear to respond to light despite being based upon cells having an anatomically fixed mesophyll conductance. Such behavior was observed in species with contrasting leaf structure using the variable J or stable isotope method of measuring $g_{m,leaf}$. A species with bifacial structure, *Arbutus* × ‘Marina’, and an isobilateral species, *Triticum durum* L., had contrasting responses of $g_{m,leaf}$ upon varying adaxial or abaxial illumination. Anatomical observations, when coupled with the proposed model of $g_{m,leaf}$ to PPFD response, successfully represented the observed gas exchange data. The theoretical and observed evidence that $g_{m,leaf}$ apparently responds to light has large implications for how $g_{m,leaf}$ values are interpreted, particularly limitation analyses, and indicates the importance of measuring $g_m$ under full light saturation. Responses of $g_{m,leaf}$ to the environment should be treated as an emergent property of a distributed 3D structure, and not solely a leaf area based phenomenon.

*Key-words*: photosynthesis; *Arbutus*; *Triticum*; internal conductance; leaf anatomy
INTRODUCTION

Mesophyll conductance to CO$_2$ ($g_m$) is understood to be the result of multiple processes within the leaf. In combination, these factors limit photosynthesis by up to 50%. Major limitations to CO$_2$ diffusion within the leaf include: diffusion in the air from the stomata to the cells, diffusion in solution in the tortuous cell wall, movement through the plasma membrane or aquaporins, diffusion through the cytosol influenced by carbonic anhydrase, and movement through the chloroplast envelopes. Most of these limitations are constant and anatomically determined (Evans et al. 2009, Nobel 1999, Terashima et al. 2011) and potentially genetically determined (Barbour et al. 2016, Jahan et al. 2014). However, the hypothesized role of aquaporins (Flexas et al. 2006, Perez-Martin et al. 2014) and carbonic anhydrase (Ho et al. 2016, Tholen & Zhu 2011) may allow the leaf to dynamically control $g_m$ in response to the environment.

A third effect that has not been routinely incorporated into the concepts of $g_m$ response to the environment is the three-dimensional nature of CO$_2$ diffusion in the leaf (Parkhurst 1986, 1994). In particular, the leaf vertical profile has varying photosynthetic capacity and cellular structure (Evans 2009, Evans and Vogelmann 2003, Ho et al. 2016, Verboven et al. 2015). Heterogeneities in the intra-leaf light absorption profiles and cell structure have previously allowed researchers to explain the difference between the light responses of leaves illuminated on the adaxial or abaxial side (Oya and Laisk 1976, Terashima 1986, Terashima et al. 1986). Similarly, irradiance of different quality, e.g. wavelength, diffuse or direct illumination, penetrates to alternative depths in the leaf (Brodersen and Vogelmann 2010, Evans and Vogelmann 2003, Terashima et al. 2009). Also, the ability of chloroplasts to move would lead to changes in the diffusion pathway (Gorton et al. 2003) as the surface of chloroplast exposed to the intercellular airspace would change (Ho et al. 2016, Tholen et al. 2008). Although little change

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in leaf $g_m$ has been observed in response to chloroplast movement (Gorton et al. 2003, Loreto et al. 2009), it is conceivable that changing between diffuse and direct light, or high versus low intensities would lead to dynamic variation in the physical basis for CO$_2$ diffusion at a local or chloroplast level. Other factors may allow a leaf to display dynamics in $g_m$, analogous to gating of water transport in aquaporins with light (Prado et al. 2013) and changing proportions of photorespiration and respiration in response to $C_i$ as light decreases (Tholen & Zhu 2011). Alternatively, Evans (2009) suggests that varying photosynthetic contributions of cells with different characteristics could lead to apparent changes in $g_m$.

The response of $g_m$ to light has not been extensively reported to date. With a few studies reporting that $g_m$ remains constant over a range of photosynthetic photon flux density (PPFD) and others showing a slight response. A response was observed from CO$_2$ response curves measured at three PPFD, where $g_m$ was higher at 1000 than at 250 µmol m$^{-2}$ s$^{-1}$ PPFD (Flexas et al. 2007). No response of $g_m$ to PPFD was found using stable isotope methods under 2% [O$_2$] (Tazoe et al. 2009). Neither study reports $g_m$ for low PPFD’s (<200 µmol m$^{-2}$ s$^{-1}$). Yin et al. (2009), measuring below 200 µmol m$^{-2}$ s$^{-1}$, observed a $g_m$ response to PPFD in wheat and account for it by fitting a phenomenological model similar to Leuning’s stomatal model (Leuning 1995) where $g_m$ is variable. Thus, it appears that there is a need to investigate the nature of $g_m$ response to light.

The following hypothesis is suggested: leaf $g_m$ responds to PPFD due to changing patterns of light penetration within the leaf leading to different contributions of each layer to bulk leaf $g_m$ ($g_{m,leaf}$). The hypothesis is further developed in the Theory section and tested using the modified variable J method, while stable isotope based $g_{m,leaf}$ values were used as confirmation of $g_{m,leaf}$ response to PPFD. Two species were chosen with bifacial or isobilateral
leaf anatomy to provide contrasting profiles of photosynthesis in leaves. As a confirmation of the
proposed model, anatomical observations were used to constrain the model, and determine if the
observed gas exchange patterns could be replicated by the model. The alternative hypotheses are
that either there is no response of \( g_{m,leaf} \) to PPFD, or that if a anatomically determined multi-layer
leaf photosynthesis model is unable to represent the observed responses, then a dynamic,
regulated mechanism of \( g_{m,leaf} \) response to light is required.

**THEORY**

A leaf can be seen as a stack of layers representing the palisade and spongy mesophyll, or a
profile (e.g. Parkhurst 1986). Layers can have very different photosynthetic properties and each
layer is influenced by the processes occurring in the adjacent layers, for example light
absorption, which will influence the CO₂ drawdown resulting from photosynthesis.

The goal of the modeling here was to reconstruct the measured net photosynthesis of a
leaf \( (A_{n,leaf}) \), PPFD absorbed by PSII \( (I_{2,j}) \) at each layer \( (j) \) and the mesophyll conductance \( (g_{m,leaf}) \)
to PPFD relationship from modelled leaf layer values \( (A_{n,j} \) and \( g_{m,j}; \) Fig. 1; see Table 1 for
definition of model variables). Applied to the extreme – a leaf with uniform cell characteristics
across layers and a constant layer based \( g_{m,j} \) – the model would form a null hypothesis (termed
null model here) to which leaves with varying layer characteristics could be compared (i.e. the
model leaf presented in Fig. 1). The general model would demonstrate if \( g_{m,leaf} \) changes with
PPFD, based upon variation in the photosynthesis of different layers in the leaf, independent of
regulated changes in the diffusion CO₂ pathway. Specifically, if layer-\( g_{m,j} \) is constant based upon
anatomy, but each layer varies in contribution to total leaf \( A_{n,leaf} \), then how does bulk leaf \( g_{m,leaf} \)
respond to PPFD?
The PPFD reaching each layer ($PPFD_j$) is:

$$PPFD_j = PPFD_{j-1} - PPFD_{abs,j-1} \quad \text{(eqn. 1)}$$

where $PPFD_{abs,j-1}$ is the PPFD absorbed by the layer before, and calculated as:

$$PPFD_{abs,j-1} = PPFD_{j-1} \alpha_{j-1} \quad \text{(eqn. 2)}$$

where $\alpha_{j-1}$ is the absorptance of the previous layer. The effective PPFD for calculating electron transport rates ($I_{2,j}$) is:

$$I_{2,j} = PPFD_j \alpha_j (1-f_j) \beta_j \quad \text{(eqn. 3)}$$

where $f_j$ is $\sim0.15$, a spectral quality correction for the relative inefficiency of white light relative to red photons (Evans 1987) and the $\beta_j$ is typically 0.5, the partitioning between PSII and I (von Caemmerer 2000). The electron transport rate for a layer of the leaf ($J_j$) is then modelled using the light-response curve equation from Ögren and Evans (1993):

$$J_j = \frac{I_{2,j} + J_{max,j}}{2 \theta_j} \left[1 - \sqrt{\left(\frac{I_{2,j} + J_{max,j}}{2 \theta_j}\right)^2 - 4 \theta_j I_{2,j} J_{max,j}}\right]$$

$$J_j = \frac{I_{2,j} + J_{max,j}}{2 \theta_j} \quad \text{(eqn. 4)}$$

where $J_{max,j}$ is the maximal electron transport rate of the $j$'th layer and $\theta_j$ is the curvature factor.

A constant intercellular airspace CO$_2$ concentration ($C_i$) is assumed across the leaf as the intercellular airspace contribution to $g_{m,leaf}$ is low for many species and leaf anatomies (Aalto & Juurola 2002, Piel et al. 2002). Although this has been debated (Parkhurst 1994), this assumption has been widely used in profile based modelling and helps to simplify the model without requiring multiple nested simulations to model the layer photosynthetic rates, and recent finite element 3D modelling has shown a mostly constant [CO$_2$] within the leaf intercellular airspace.
(Ho et al. 2016). However, the chloroplastic CO$_2$ concentration ($C_{c,j}$) for each layer of the leaf has to be found using an numerical solution to obtain the gradient for CO$_2$ diffusion necessary to compute $A_{n,j}$. This is done by minimizing the difference between eqn. 5 and 6 for each layer of the leaf:

$$A_{n,j} = g_{m,j}(C_i - C_{c,j}) \quad \text{(eqn. 5)}$$

and

$$A_{n,j} = (1 - \Gamma^*_{j}/C_{c,j}) \times \min(V_{c,Rubisco,j}, V_{c,RuBP,j}) - R_{d,j} \quad \text{(eqn. 6)}$$

where:

$$V_{c,Rubisco,j} = C_{c,j}V_{c,max,j}/(C_{c,j} + K_{c,j}(1 + O/K_{o,j})) \quad \text{(eqn. 7)}$$

$$V_{c,RuBP,j} = J_j/(4 + 8\Gamma^*_{j}/C_{c,j}) \quad \text{(eqn. 8)}$$

and $R_{d,j}$ is the mitochondrial respiration in the light ($\sum_{j=1}^{l} R_{d,j} = R_{d,leaf}$), $\Gamma^*_{j}$ is the photosynthetic compensation point in the absence of respiration (same for each layer), $V_{c,Rubisco,j}$ the carboxylation rate under Rubisco-limited conditions, and $V_{c,RuBP,j}$ the carboxylation rate under RuBP-limited conditions. The numerical solution is found for each layer $j$. In eqn. 5, $g_{m,j}$ is a fixed value specific for each layer $j$.

The leaf mesophyll conductance can then be calculated based upon a constant $g_{m,j}$ at each layer and the weighted contribution of each layer to leaf $A_{n}$. This principle can be illustrated by the situation where, as a layer of the leaf tends towards zero net photosynthesis, then the layer contributes a decreasing signal to the measured $g_{m}$ of the whole leaf. An assimilation-weighted
\[ g_m,\text{leaf} = \frac{A_{n,\text{leaf}} \cdot \Sigma_{j=1}^{l} A_{nj} (C_{ij} - C_{cj})}{C_{i} - \frac{\Sigma_{j=1}^{l} A_{nj} C_{cj}}{A_{n,\text{leaf}}}} \]  

(eqn. 9)

A useful aspect of the model is that it can be used to simulate the \( g_m \) light response of a leaf regardless of the direction of illumination. This modelling approach is used for a leaf with three layers (Fig. 1), for instance two palisade and one spongy mesophyll layers (increasing the number of layers produces a similar response; data not shown). The theoretical leaf presented in Figure 1 was parameterized assuming all layers had the same photosynthetic parameters, termed the null model, i.e. \( g_{m,j} \) was equal for each layer and structurally fixed (parameters for the null model are given in the Figure 1 legend). That is, layer based values of \( g_{m,j} \) did not vary in the null model, but the contribution of each layer to the whole leaf \( g_{m,\text{leaf}} \) signal did vary based upon layer specific photosynthetic rates. At high PPFD all layers contributed a similar photosynthesis rate, and thus \( g_{m,\text{leaf}} \) was roughly the additive values of \( g_{m,j} \) (Fig. 1; eqn. 9). Under the lowest PPFD the adaxial layer contributed a higher photosynthesis signal, resulting in that layer contributing the most to \( g_{m,\text{leaf}} \) while the abaxial layer respired resulting in a negative contribution to \( g_{m,\text{leaf}} \) and a low total \( g_{m,\text{leaf}} \) value. When the leaf respires, i.e. \( A_{n,\text{leaf}} \) is negative, then the \( g_{m,\text{leaf}} \) must increase back to the maximum value as the flux and gradient would be inverted (these respiratory responses are at the very left most side of the \( g_{m,\text{leaf}} \) PPFD response in Fig. 1).

However, a caveat of eqn. 9 is that when \( A_{n,\text{leaf}} \) equals zero then the denominator, the gradient \((C_i - C_c)\), is zero, which makes the estimation of \( g_{m,\text{leaf}} \) undefined. This is an unavoidable and inherent issue in the common conception of \( g_m \) in the form of:
\[ g_m = \frac{A_n}{(C_i - C_c)} \]  
(eqn. 10)

when \( C_c \) equals \( C_i \) (see also eqn. 11). Considering this, estimating \( g_{m,leaf} \) under low light conditions leading \( A_{n,leaf} \) to equal zero is unreliable. This is reached under very low light intensities, below 50 \( \mu \text{mol m}^{-2} \text{ s}^{-1} \) PPFD in our null model of a leaf (Fig. 1). Nonetheless, in a real leaf, this mathematical issue is simply not present and \( C_i \) must equal \( C_c \) at least twice a day. In that case, \( g_m \) as a physical resistance still exists, but would be unmeasurable.

Thus, it is clear from a theoretical standpoint that \( g_{m,leaf} \) must appear to respond to light in a structural manner, consistent with pools of cells contributing to photosynthesis differentially under varying light. Evans (2009) predicted \( g_{m,leaf} \) to respond to PPFD in this manner, while Parkhurst (1994) and Lloyd et al. (1992) suggested that differing photosynthetic contributions of cells would lead to apparent changes in \( g_{m,leaf} \). The rest of this paper is directed to answering:

1) Is such behavior observed in leaves?

2) Are the observed light responses of \( g_{m,leaf} \) consistent with constant anatomical characteristics and varying structural (layer) contributions to photosynthesis, or are dynamic, regulated processes necessary to explain the observed light responses?

3) What broader implications does this model of the \( g_{m,leaf} \) response to PPFD have for contrasting leaf anatomies, and for the measurement of \( g_m \)?

**MATERIALS AND METHODS**

**Species used and plant growth conditions**
Arbutus × ‘Marina’ (Arbutus unedo L. × A. andrachne L.) year-old saplings were used to represent a leaf with bifacial anatomy and high leaf mass area (LMA: 97.36 g m⁻²). Triticum durum L. cv. Kronos two-month-old seedlings were used to represent a leaf with high photosynthetic capacity and approximately isobilateral anatomy (LMA: 12.34 g m⁻²).

Arbutus plants were grown outdoors during the fall at the UC Davis Arboretum Nursery in 4 L pots and were transferred to an environmentally-controlled greenhouse for approximately two weeks to acclimate. Temperature was 25/18°C (day/night) and maximal PPFD from sunlight was ~800 µmol m⁻² s⁻¹. Triticum seeds were sown in 4 L pots filled with a coarse substrate (1/3 sand, 1/3 peat, 1/3 redwood compost) in the same greenhouse. Both species were fertigated daily when irrigated.

The flag leaves of Triticum were measured; these leaves are positioned at a high angle with illumination occurring from either side, or may even ‘flip’ presenting the abaxial surface upwards. For Arbutus, a fully expanded leaf, with a plastochron index of five to seven, were measured.

Gas exchange: variable J method

Plants were transferred to the lab into a custom-made cabinet (~1.2 m³), allowing for the control of temperature (maintained at 25±1°C), vapor pressure deficit (VPD; 1.5±0.2 kPa), and PPFD (set at 800 µmol m⁻² s⁻¹ at the top of the plant), and air mixing. A custom gas exchange chamber was used to maximize resolution under low photosynthesis. Leaf gas exchange was measured inside the growth cabinet using a LI-6400XT with a 2x3cm clear top PAM−2000 adaptor chamber (LI-COR Biosciences, Lincoln, NE, USA) and equipped with a PAM−2000 chlorophyll fluorometer (Heinz Walz GmbH, Effeltrich, Germany). PPFD was provided to the leaf by a
white light (LI-COR6400-18A) providing equal quantities of red, blue and green PPFD, and was placed ~2 cm above the leaf surface, with the angle adjusted to avoid the PAM-2000 probe shading the leaf, whilst still illuminating the leaf homogeneously, similar to Bellasio and Griffiths (2014). The chamber was covered with black cloth to shade the leaf from external light. The leaf was allowed to stabilize with the adaxial side being illuminated at leaf chamber CO$_2$ mole fraction ($C_a$) of 380 µmol mol$^{-1}$, PPFD of 1000 µmol m$^{-2}$ s$^{-1}$, VPD of 1.5±0.1 kPa, flow of 500 µmol s$^{-1}$, and leaf temperature of 25±0.2°C. Gas exchange was recorded simultaneously with steady state chlorophyll fluorescence under light ($F_s$) and maximum fluorescence under saturating light ($F_{m'}$; ~15000 µmol m$^{-2}$ s$^{-1}$), used to compute the photochemical efficiency of PSII ($\Phi_{PSII} = (F_{m'} - F_s) / F_{m'}$).

Following the measurement under ambient $C_a$, [CO$_2$] was increased to lower the photorespiratory bias on $g_m$ estimates when evaluated at low $C_i$ (Tholen & Zhu 2011). A normalized $C_i$ of 280 µmol mol$^{-1}$ was used, which is the minimal $C_i$ at which $g_m$ plateaus; see Théroux-Rancourt et al. (2014). A light response curve using adaxial illumination was measured at normalized $C_i$ at PPFD of 950, 500, 350, 230, 140, 85, and 45 µmol m$^{-2}$ s$^{-1}$. After the last point, the leaf was inverted and equilibrated at 800 µmol m$^{-2}$ s$^{-1}$ PPFD for at least one hour. The leaf was then equilibrated to 950 µmol m$^{-2}$ s$^{-1}$ PPFD before measuring the same light response as above. The lights were then turned off and the plant was shaded with a black cloth, and dark respiration ($R_n$) was measured after ~20 min. The order in which the sides of the leaves were measured did not affect the light responses (data not shown), and so the adaxial side was chosen as the first side measured.

Mesophyll conductance to CO$_2$ was estimated using the variable $J$ method (Harley et al. 1992):
where $J_f$ is the photochemical electron transport rate estimated from chlorophyll fluorescence, and $\Gamma^*$ is assumed to be 37.4 µmol mol$^{-1}$ (Bernacchi et al. 2002). The use of a normalized and constant $C_i$ for the estimation of $g_m$ limits the potential bias caused by an inaccurate $\Gamma^*$, which means that the observed response is less sensitive to this issue, as seen in the initial variable $J$ method paper (Harley et al. 1992) and in one of our previous reports (Théroux-Rancourt et al. 2014). Dark respiration was used as a rapid proxy for $R_{d,leaf}$, considered to be half of $R_n$ (Niinemets et al. 2009, Théroux-Rancourt et al. 2014). The electron transport rate estimated from chlorophyll fluorescence was calibrated according to the following linear relationship (Hassiotou et al. 2009):

$$J_f = \left[ (\Phi_{PSII} s) + c \right] PPFD \tag{eqn. 12}$$

where $s$ represents the ratio of $\Phi_{CO2}$ (the gas exchange-based photochemical quantum yield) to $\Phi_{PSII}$, and $c$ represents the intercept of the relationship between $\Phi_{PSII}$ and $\Phi_{CO2}$. Calibration was performed under ambient, 21% $O_2$ conditions following the method described in Théroux-Rancourt et al. (2014). Using detailed $A_n$-$C_i$ curve analysis combined with chlorophyll fluorescence, $s$ was fit using the RuBP-limited version of eqn. 3 of Éthier et al. (2006). This method hence solves $s$ (and so $J_f$) and $g_m$ simultaneously using the measured $A_n$-$C_i$, $\Phi_{PSII}$, and $R_d$. The $s$ values estimated were between 0.319 and 0.42 for *Arbutus* and between 0.39 and 0.43 for *Triticum*, and $c$ values were between $1 \times 10^{-4}$ and 0 for *Arbutus* and between 0.01 and 0 for *Triticum*. 

$$g_{m,leaf} = \frac{A_{n,leaf}}{C_i - \frac{\Gamma^* J_f + \frac{3 (A_{n,leaf} + R_{d,leaf})}{J_f}}{4 (A_{n,leaf} + R_{d,leaf})}} \tag{eqn. 11}$$
Gas exchange: stable isotope method

On a separate set of plants from those above, light response curves were performed as above, but the air exiting the LI6400XT cuvette was collected and analyzed for stable isotope composition. A three-way valve was added to the chamber exhaust tube, the third port connected to a ~6 m sampling tube (> 50 cm³). After measuring \( g_{\text{m,leaf}} \) as above, but under a flow of 200 µmol s\(^{-1}\) to maximize CO\(_2\) drawdown within the cuvette and at PPFD of 950, 700, 450, 350, 230, 120, 80, or 45 µmol m\(^{-2}\) s\(^{-1}\), the valve was opened toward the sampling tube and air was allowed to flow for ~5 min, allowing > 20 times air change within the tube. To sample air, the valve was returned to its original direction along the cuvette exhaust route, and 20 cm³ air was slowly sucked from the tube into a gas-tight glass syringe through a brass luer-lock fitting. The syringe’s valve was closed, a needle connected, and the needle was flushed with some of sampled air before injecting 12 ml of air into a vial (10 ml Exetainer, Labco, UK). Air was sampled at different PPFD for *Triticum* (950, 450, 230, and 80 µmol m\(^{-2}\) s\(^{-1}\)) and *Arbutus* (950, 700, 350, and 120 µmol m\(^{-2}\) s\(^{-1}\)) as the latter closed stomata rapidly at low PPFD and could not be left for over 10 min below 100 µmol m\(^{-2}\) s\(^{-1}\) PPFD. Air was then sampled from an empty cuvette to measure the isotopic signature of the incoming air at the same chamber [CO\(_2\)] (\( C_s \)) as the samples measured above ([CO\(_2\)] varied due to normalizing \( C_i \)) in order to get the reference carbon isotopic composition of the tank CO\(_2\) (~ -36‰).

Carbon isotopic composition (\( \delta^{13}C \)) of the air samples was measured within one week of sampling at the UC Davis Stable Isotope Facility using a ThermoScientific PreCon-GasBench system interfaced to a ThermoScientific Delta V Plus isotope ratio mass spectrometer (ThermoScientific, Bremen, Germany). CO\(_2\) was sampled by a six-port rotary valve (Valco, Houston, TX) with an 100µL loop programmed to switch at the maximum CO\(_2\) concentration in
the helium carrier gas. The CO\textsubscript{2} was then separated from N\textsubscript{2}O and other residual gases by a Poroplot Q GC column (25m x 0.32mm ID, 45°C, 2.5 mL/min). A pure reference gas (CO\textsubscript{2}) was used to calculate provisional $\delta^{13}$C values. Final $\delta^{13}$C values were obtained by correction to $\delta^{13}$C values for laboratory standards (calibrated directly against NIST 8545).

Mesophyll conductance was estimated from photosynthetic parameters and the carbon isotopic discrimination against $^{13}$CO\textsubscript{2} ($\Delta^{13}$C), accounting for the ternary effect (Farquhar & Cernusak 2012) using the equations of Evans and von Caemmerer (2013). $\Delta^{13}$C was computed as (Evans et al. 1986):

$$\Delta = \frac{1000\xi(\delta^{13}C_{sam} - \delta^{13}C_{ref})}{1000+\delta^{13}C_{sam} - \xi(\delta^{13}C_{sam} - \delta^{13}C_{ref})}$$  

(eqn. 13)

where $\delta^{13}C_{sam}$ and $\delta^{13}C_{ref}$ are the isotopic compositions of the LI6400XT cuvette air with and without a leaf, and $\xi = C_a / (C_a - C_s)$. The value of $\xi$ ranged on average from 8 under 950 µmol m\textsuperscript{-2} s\textsuperscript{-1} PPFD to 50 under 100 µmol m\textsuperscript{-2} s\textsuperscript{-1} PPFD.

Microscopy

Leaves were prepared for microscopy using methods from Bozzola and Russell (1992), and Russin and Trivett (2001). Leaves were fixed in Karnovsky’s fixative. Tissues were rinsed with 0.1M PO\textsubscript{4} buffer and post-fixed for 2 h in 1% buffered osmium tetroxide. Leaves were dehydrated with ascending concentrations of ethyl alcohol with three changes at 100%, transitioned 1:1 with propylene oxide, and dehydrated using two changes of pure propylene oxide. Infiltration began using Epon/Spurr’s resin in three ascending concentrations with propylene oxide. Finally, three changes of resin with microwave assistance were done before
overnight polymerization in capsules. For light microscopy, semi-thin sections were cut using a Leica Ultracut UCT ultramicrotome and were stained with 2% Methylene Blue/Azure II before being observed at 40× magnification with a Axio Imager A2 microscope (Zeiss, Oberkochen, Germany). For transmission electron microscopy, ultrathin sections were cut using a Diatome diamond knife and picked up on 150 mesh copper grids. The sections were stained with uranyl acetate and lead citrate before viewing with a Phillips CM120 Biotwin (FEI, Hillsboro, OR) and equipped with a Gatan MegaScan 794/20 camera.

For each species, two different leaves were analyzed, with two to three cross sections per leaf, a total of five cross sections per species. Structural traits were analysed using the ImageJ software (Schneider et al. 2012). Leaf mesophyll thickness (t) was the average distance between the ad- and abaxial epidermis, and divided to create three artificial layers of equal thickness parallel to the epidermises. For the whole leaf and within each layer, the total area of mesophyll and intercellular airspace (IAS) were measured to estimate the leaf mesophyll porosity: \( f_{\text{IAS}} = \frac{\text{area IAS} \; (\mu m^2)}{\text{mesophyll area} \; (\mu m^2)} \). The length of mesophyll cell wall exposed to the IAS \( (L_m; \; \mu m) \) was measured in order to compute the surface area of mesophyll exposed to the IAS per leaf area \( (S_m) \) as:

\[
S_m = \frac{L_m}{W} \times F
\]  

(eqn. 14)

where \( W \) is the width of the section and \( F \) is the curvature correction factor to convert measured length into surfaces. This correction factor was computed following Thain (1983) by measuring, for each layer, the major and minor axes of at least ten cells for two different leaves (\( F \) values are shown in Table 2). The whole leaf \( S_m \) was computed from the whole leaf average \( F \).
For electron micrographs, the thickness of the cell wall \((T_{cw})\) and cytosol \((T_{cyt})\) was measured (> five different cells). The cell wall length exposed to the IAS of individual cells was measured \((L_m)\), the length of chloroplast exposed to the IAS \((L_c; \mu m)\), and the surface area of chloroplasts exposed to the IAS per leaf area \((S_c)\) equaled:

\[
S_c = S_m \times \frac{L_c}{L_m}.
\]  
(eqn. 15)

\(S_c\) was computed for each layer, and the whole leaf \(S_c\) was computed from the whole leaf \(S_m\) multiplied by the whole leaf average chloroplast coverage of mesophyll cells.

The liquid phase resistance \((r'_{liq})\) to CO\(_2\) was estimated as the sum of all the liquid phase components in the diffusion path from the cell wall to the stroma using the equations of Evans et al. (2009) adjusting anatomy specific lengths. Liquid phase resistance on a chloroplast surface area basis was converted to a leaf area basis to provide an anatomical estimate of \(g_m\) using:

\[
g_m = \left(\frac{1}{r'_{liq}}\right)S_c.
\]  
(eqn. 16)

**Simulations**

The model described in the *Theory* section was used to simulate the sensitivity of \(g_{m,leaf}\) to PPFD responses for a two-layer leaf, to vary parameters singly, or used to fit a three-layer model to the mean observed data for the two species. To constrain fitting, the minimum sum of squares was found using the L-BFGS-R constrained optimization algorithm with the *optim* function of R (version 3.3.1). Boundary conditions were set to limit the different parameters within a relevant range. This consisted of limiting the layer-maximum value to the whole leaf maximum or anatomical data \((J_{max,j}: \text{the whole leaf value fitted from gas exchange and chlorophyll} \ldots\))
fluorescence; \( g_{m,j} \): maximum layer-specific anatomical value; \( \alpha_j \): the maximum predicted layer-specific \( \alpha + 10\% \), generated according to the relative \( S_c \) profile constrained to yield a total leaf absorption of 0.85 (layer-specific \( \alpha \) values represent the fraction of light absorbed within that layer – generated using equations 1 and 2 – and not the fraction of the total incoming light)). For \( R_{d,j} \), a high upper boundary was set to (5 \( \mu \text{mol m}^{-2} \text{s}^{-1} \)), and \( \theta_j \) was constrained between 0.0001 and 0.9999.

The relative anatomically based relationship between each layer was also included in the fitting procedure so that the resulting parameters would follow a profile similar to the observed anatomy. Specifically, fitting SSE was penalized if \( J_{\text{max},i} \) and \( R_{d,i} \) did not follow the relative profile in \( S_c \), while \( g_{m,i} \) followed the profile in \( g_m \) estimated from anatomy (Table 2). This allowed for different absorption profiles depending on the illuminated surface, while providing a unique value for each layer. This value corresponded to the absorbed fraction of incoming light at layer \( j \), thus taking into account the absorbed fraction at layer \( j-1 \). Total leaf absorption was constrained to 0.85, a value commonly used in the literature (e.g. Evans 2009).

The sum of squares to be minimized was the sum of nine constraints, i.e. the squared differences between: i) \( A_{n,\text{leaf}} \) observed and predicted by the model for ad- and abaxial illumination, ii) \( g_{m,\text{leaf}} \) observed and predicted for ad- and abaxial illumination, iii) the \( C_{c,j} \) values input and those predicted by using the result of eqn. 6 and calculating \( C_{c,j} \) from rearranging eqn. 5, iv) predicted whole leaf absorptance and a default value of 0.85, v) the observed and predicted initial slope of the \( A_{n,\text{leaf}} \) to PPFD response for ad- and abaxial illumination, and vi)-ix) the sum of the squared difference between the reference relative profile and the fitted relative profile for \( J_{\text{max},j} \), \( g_{m,j} \), \( R_{d,j} \), and \( \alpha_j \). Each constraint was weighted so that the resultant sum of squares for each was within one order of magnitude of the others.
The optimization was ran on over 500 sets of starting values that followed the relative profiles in $S_c$ ($J_{\text{max},j}$, $g_{m,j}$, $R_{d,j}$) or values from the literature ($a_j$), and starting values for $\theta_j$ were selected randomly between 0.001 and 0.999. Solutions within 10% of the SSE of the best fit were selected. For $a_j$, we validated the fit gradient based upon measurements of $S_c$ profiles by contrasting it to the published absorption profiles of a bifacial leaf (spinach; Evans and Vogelmann 2003) and an isobilateral leaf (eucalyptus; Evans and Vogelmann 2006) to represent Arbutus and Triticum, respectively.

RESULTS

Response of net photosynthesis and mesophyll conductance to PPFD

The bulk net photosynthesis of Triticum and Arbutus were typical of reported responses to irradiance from adaxial illumination (Fig. 2). However, leaf mesophyll conductance varied approximately proportionally with net photosynthesis at lower PPFD (Fig. 2). A switch from adaxial to abaxial illumination had similar effects on the $A_{\text{n,leaf}}$ and $g_{m,\text{leaf}}$ of Arbutus with a ~40% decrease. In contrast, the isobilateral leafed species, Triticum, had similar photosynthesis regardless of the side of illumination, but a large decrease in $g_{m,\text{leaf}}$ upon abaxial illumination. The $C_i$ was kept constant at 280 $\mu$mol mol$^{-1}$ during the PPFD responses. Thus, the observed responses are not due to an apparent response of $g_m$ to CO$_2$ concentration.

The stable isotope method of estimating $g_{m,\text{leaf}}$ was used to establish whether the variable $J$ chlorophyll fluorescence based method could have biases that led to the response of $g_{m,\text{leaf}}$ to PPFD. Response curves measured simultaneously using the chlorophyll fluorescence and stable isotope methods showed similar responses of $g_{m,\text{leaf}}$ to PPFD (Fig. 3). The estimated values of
Using both methods were directly proportional to each other, indicating that the curved \( g_{m, \text{leaf}} \) response to PPFD was not due to a bias by the chlorophyll fluorescence method. However, as is typical in the literature (e.g. Vrábl et al. 2009), the two methods had quantitatively different absolute values.

**FIGURES 2 AND 3 COULD BE PLACED HERE**

**Anatomical description of leaves and layer specific structural parameters**

Dividing the leaves into three equal-depth layers led to distinct profiles of anatomy in both bifacial *Arbutus* and isobilateral *Triticum* (Fig. 4; Table 2). For *Arbutus*, this clearly separated a dense palisade layer of one cell of \( \sim 80 \mu \text{m} \) length adjacent to the adaxial epidermis, followed by a loose palisade of two blunt cylindrical cell layers, and four-layers of loosely-packed spheroid to cylindrical cells in the spongy mesophyll adjacent to the abaxial epidermis, where all stomata were located (i.e. a hypostomatous leaf) (see Table 2). *Triticum* showed a more symmetrical profile from adaxial to abaxial epidermis, with a layer of one cylindrical palisade cell of \( \sim 50 \mu \text{m} \) length touching both epidermises, and a spongy-like middle layer composed of spheroid and invaginated cells typical of grasses (e.g. Giuliani et al. 2013). The abaxial epidermis was more porous because of the presence of more stomata per section width, hence more substomatal cavities which increase the fraction of intercellular airspace.

**TABLE 2 AND FIGURE 4 COULD BE PLACED HERE.**

*Arbutus* leaves were thicker and had higher mesophyll cell wall area \( (S_m) \) and chloroplast area \( (S_c) \) per unit leaf area than *Triticum* (Table 2), although *Triticum* showed a higher exposed surface on a mesophyll volume basis \( (0.14 \text{ (Triticum) vs. } 0.10 \text{ (Arbutus)} \mu \text{m}^{-1}; \text{ see Nelson et al.} \)
2005 about the use of this metric). *Arbutus* had similar values of $S_m$ and $S_c$ in the adaxial and middle layers, and these values decreased in the abaxial layer. *Triticum* had high values on both the ad- and abaxial surfaces and lower values in the center (Table 2). A similar pattern was found in cell wall thickness ($T_{cw}$), *Arbutus* had an increasing gradient and *Triticum* a low-high-low pattern. Cytoplasm thickness ($T_{cyt}$) varied less between cell layers, and *Triticum* had a ~three times smaller $T_{cyt}$ than *Arbutus*.

**Layer-based modelling of leaf photosynthesis and mesophyll conductance**

A three-layer model (*Theory*) was able to represent the observed gas exchange data for either species, including the response of $g_{m,leaf}$ to ad- or abaxial illumination (Fig. 2). In the model, the $g_{m,j}$ values for each layer remained constant, while the weighted values vary resulting in changing $g_{m,leaf}$ (estimated from eqn. 9). The resulting whole leaf values fit the measured $A_{n,leaf}$ values for both species relatively well (Fig. 2). The measured and predicted values for *Arbutus* were similar for the three-layer model, and the range of the best solutions was narrow. The range of the solutions was wider for *Triticum* and there was disagreement between the fitted and measured $g_{m,leaf}$ data in the range of about 125 to 500 µmol m$^{-2}$ s$^{-1}$ PPFD.

The cause of the varying contribution of each layer to $g_{m,leaf}$ was due to changing layer-specific $A_n$ values (Fig. 5). Only the directly illuminated layers approached saturation, while all other layers had considerably lower photosynthesis, decreasing the $g_{m,j}$ signal from those layers. For this reason, the $g_{m,leaf}$ response was considerable when PPFD was lower than 500 µmol m$^{-2}$ s$^{-1}$. The low photosynthesis in the adaxial and middle layers in *Arbutus* with abaxial illumination led to the lower $g_{m,leaf}$ than expected from adaxial illumination response (Fig. 2 and 5).
To parameterize the model, the parameters were fit to the gas exchange data, but constrained by the observed relative anatomical profiles for a number of characteristics. Thus, *Triticum* presented an even $g_m$ profile throughout the leaf from the adaxial to the abaxial epidermis (average values of 0.07 mol m$^{-2}$ s$^{-1}$ for each layer; Fig. 6), which followed the relative $S_c$ profile and anatomically based $g_m$ profile (Table 2). *Arbutus* also followed the relative profile in $S_c$ and anatomically based $g_m$ with average $g_{m,j}$ values of 0.06, 0.06, and 0.03 mol m$^{-2}$ s$^{-1}$ from ad- to abaxial epidermis. The fitting of a relative profile was necessary, as the anatomical correlate of $J_{\text{max}}$ and $R_d$, $S_c$, is not scaled, and the anatomical estimate of $g_m$ is likely to be an overestimate, accounting for only idealized conditions.

$J_{\text{max},j}$ and $R_{d,j}$ were similarly constrained by the $S_c$ relative profile (Fig. 6, Table 2). Layer-specific $g_m$ values presented a very narrow range, as were $R_{d,j}$ values, whereas $J_{\text{max},j}$ had greater variation, but still resulted in a similar whole leaf value (Fig. 2) to layer-specific light response curves (Fig. 5). The leaf based parameter values were additive values for $J_{\text{max},j}$ and $g_{m,j}$ representing a value close to the whole leaf measured ($g_m$) or estimated value ($J_{\text{max}}$). For the sum of $R_{d,j}$, this was a somewhat high value compared to what was measured (between 0.5 and 1.2 $\mu$mol m$^{-2}$ s$^{-1}$).

The light response curvature ($\theta_j$) exhibited the most variation within each layer as it was not constrained to a specific profile, but the median of each layer was above 0.70 for each species (Fig. 6). Layer specific absorptance ($\alpha_j$) was relatively constrained within each species, displaying little variation between fits (Fig. 6). In order to compare the fitted data with Evans and Vogelmann’s (2003, 2006) data, absorption profiles were produced for ad- and abaxial illumination (Fig. 7). The value for each layer was computed using equations 1 and 2, and the resulting layer-specific PPFD value was divided by sum of the PPFD absorbed. For Evans and
Vogelmann’s data, the absorption profile was split into three equal thickness layers for both blue and green light. From these profiles, the median fits for *Triticum* were consistent with the observed data for the isobilateral *Eucalyptus*, regardless of direction of illumination. In contrast, for *Arbutus*, the light absorption profiles predicted from the fit, were only similar to bifacial spinach for abaxial illumination (Fig. 7).

FIGURES 5, 6 AND 7 COULD BE PLACED HERE.

To investigate the impact of each parameter on the $g_{m,\text{leaf}}$ to PPFD response, the results of the layer-based model were tested for sensitivity of the $g_{m,\text{leaf}}$ to PPFD relationship to parameter values for each layer using a minimal two layer model. At a low PPFD, a 50% decrease in the parameter value of the lower layer, i.e. the layer the furthest away from the light source, resulted in most sensitivity for $\alpha_2$ ($\alpha_1$ of the lower, second layer), then $R_{d,2}$ and little sensitivity for the $\theta_2$, $J_{\text{max},2}$ or $g_{m,2}$ (Fig. 8). But at high PPFD all parameters had effects on the $g_{m,\text{leaf}}$ to PPFD response, in the following order, from the greatest to the smallest effect: $g_{m,2}$, $\alpha_2$, $J_{\text{max},2}$, $\theta_2$, $R_{d,2}$.

As the estimation of $g_{m,\text{leaf}}$ is undefined at $A_{n,\text{leaf}}$ values close to zero (between -0.1 and 0.1 µmol m$^{-2}$ s$^{-1}$), these values were removed from Fig. 8. Yet, a high and ‘jumpy’ sensitivity to parameter values at very low PPFD remains and was due to $g_{m,\text{leaf}}$ being affected by CO$_2$ loss out of the leaf and calculation from low fluxes in eqn. 9 (a low denominator; see the Theory section).

FIGURE 8 COULD BE PLACED HERE.

DISCUSSION

Structural variation in photosynthesis across the leaf explains the $g_{m,\text{leaf}}$ response to PPFD
Two species of plants with contrasting leaf anatomy both displayed large decreases in the total mesophyll conductance to CO₂ using the modified variable $J$ method (Fig. 2) or stable isotopes (Fig. 3). The modelling of layer-based photosynthesis within the leaf could account for the observed responses of $g_{m,\text{leaf}}$ to PPFD despite layer-based $g_{m,j}$ being held constant (Fig. 5 and 6).

From the results and theory above, it appears that even if $g_m$ per leaf layer is constant and structurally based then this can result in apparently dynamic responses of $g_{m,\text{leaf}}$ to light. Mesophyll conductance response to PPFD has not been investigated in detail in the literature before, to our knowledge, although Evans (2009) suggested the principle upon which the modelling is based. The current observed response of $g_{m,\text{leaf}}$ to PPFD is difficult to compare to past reports. Firstly, the most dramatic decrease in $g_{m,\text{leaf}}$ occurs at the lowest PPFD’s, while past responses were measured at PPFD’s of greater than 200 µmol m⁻² s⁻¹ (Flexas et al. 2007, Tazoe et al. 2009, Yamori et al. 2010). Indeed, the null model (Fig. 1) had only a ~30% decrease in $g_{m,\text{leaf}}$ at a PPFD of 150 µmol m⁻² s⁻¹, but rapidly dropped near zero at lower PPFD. Yin et al. (2009) did measure below 200 µmol m⁻² s⁻¹, but did not keep $C_i$ constant. Indeed, if $C_i$ is variable during the measurement of the PPFD response there may be a compensatory effect. That is, at higher $C_i$’s, which occur at low light, $g_{m,\text{leaf}}$ may increase due to the positive CO₂ effect at low $C_i$’s (Tholen & Zhu 2011). The results here were partially de-trended of this effect as $C_i$ was kept constant. Previous experiments did not maintain $C_i$ constant when measuring light responses (Flexas et al. 2007, Tazoe et al. 2009, Yamori et al. 2010).

Anatomical profiles predict the observed gas exchange responses
The model proposed here was able to accurately represent the observed gas exchange data, when parameterized by the relative profiles of the species anatomical features ($S_c$ and $g_m$). The observed $g_m$ to PPFD responses are consistent with the observed anatomy. Thus, the modelling results can be used to investigate within-leaf profiles in photosynthetic parameters ($J_{\text{max}}$, $g_m$, and $R_d$). Interestingly, the various solutions for *Arbutus* and *Triticum* were in very narrow range of parameters for parameters like $g_{m,j}$, $R_{d,j}$, and $\alpha_j$, while more variation in absolute values were fit for $J_{\text{max},j}$ and $\theta_j$ (Fig. 6). However, *Triticum* exhibited more variation than *Arbutus*, where $g_{m,j}$, $\theta_j$, and $J_{\text{max},j}$ showed a few outlier groups.

The only parameter that somewhat diverges from the expected profile is $\alpha_j$, the layer specific absorptance. For *Triticum*, the light absorption profile was close to what was expected from the isobilateral *Eucalyptus pauciflora* using $^{14}$C assimilation profiles (Fig. 7; Evans and Vogelmann 2006). As leaves of *Triticum* can easily move with wind, the absorptance profile may reflect the ability to take full advantage of the light independent of the illuminated surface. For *Arbutus*, the abaxial layer absorbs a higher amount of light which could be explained by the increased scattering due to the more random distribution of the cells in the spongy mesophyll, as shown by Evans and Vogelmann (2003). However, the adaxial and middle layers have lower values than what would be expected from the bifacial spinach of Evans and Vogelmann (2003). Although the $\alpha_j$ values for those two layers could be underestimated compared to Evans and Vogelmann (2003) and to $S_c$, a proxy for chloroplast volume, the values do follow an anatomical trend as those two layers consist mainly of palisade mesophyll known to channel light to the deeper layers of the leaf (Vogelmann 1993). Nonetheless, the important differences in the fitted light absorption profile for adaxial illumination and the profile for Evans and Vogelmann (2003) data are a result of the simulation procedure, for which the abaxial illumination seemed to weigh
more on the final parameter values than the photosynthesis data measured with adaxial illumination. Hence, the predicted photosynthetic parameter values for each layer, based upon the model, are not necessarily robust even if they are mostly in a range expected based upon anatomy, but do give an indication of the layer based parameters required to result in the observed responses.

Limitations to this analysis

The modelling conducted here does not account for variation in CO₂ diffusion pathways through the mitochondrion due to photorespiration and respiration (Tholen & Zhu 2011). Thus at very low light (<~25 µmol m⁻² s⁻¹) or [CO₂], conditions where respiratory CO₂ evolution is high, the current model will over predict values for $g_{m,leaf}$. These light conditions were not investigated in the current work, and thus should not affect the current results, but these effects may be usefully incorporated in future investigations.

The variable $J$ method has known sensitivity to calibration conditions (Gilbert et al. 2012) and is based upon chlorophyll fluorescence accurately estimating the quantum yield of the leaf profile (Evans 2009). The former issue was dealt with by calibrating the variable $J$ method using the modified calibration approach of Théroux-Rancourt et al. (2014), moderate to low PPFD’s and by reducing noise through the use of a three times larger leaf area than is standard for fluorescence with the LI-6400XT. The potential for a gradient in chlorophyll fluorescence signal from within the leaf does pose a problem for the variable $J$ method. The discrepancies between the measured and fitted responses in the 125 to 500 µmol m⁻² s⁻¹ PPFD region can arise from a potential artifact from the chlorophyll fluorescence measurement. As pointed out by
Oguchi *et al.* (2011), chlorophyll fluorescence using a PAM fluorometer (a Walz PAM 101 in their case) leads to the measurement of a weighted signal corresponding to ~200 to 250 µm below the epidermis. Hence, when under PPFD that saturates one layer but not the deeper layers, the resulting signal might potentially be biased towards higher Φ_{PSII} values than the entire leaf, causing measurement errors in the variable \( J \) method. However, the stable isotope method is robust to this issue, as it does not rely on chlorophyll fluorescence. Stable isotopes confirmed the observed drop in \( g_{m,leaf} \) at low PPFD. Given the mathematical proof that \( g_{m,leaf} \) should respond to PPFD, it can be expected that regardless of any methodological limitations, that bulk \( g_m \) apparently varies with light.

Another limitation of this study could come from chloroplast movement. Mesophyll conductance increases as a consequence of the increase in \( S_c \) when chloroplasts move from the profile to the face position, as would happen in a shift from strong white light to blue-filtered light (*Tholen et al.* 2008). As blue light is mainly absorbed close to the illuminated surface, chloroplasts deeper in the leaf would migrate to the cell faces, implying that \( S_c \) would dynamically increase in the middle and furthest layer from the light source. Potentially this effect would allow \( g_{m,j} \) to increase and slightly increase the apparent \( g_{m,leaf} \) signal. It would be interesting to target chloroplast movement as a regulated light response of \( g_m \).

**General \( g_{m,leaf} \) responses to PPFD for diverse leaves**

General predictions can be made on how the \( g_{m,leaf} \) of diverse anatomies will respond to light. Leaves that do not saturate photosynthesis are likely to have \( g_{m,leaf} \) that responds to PPFD (null model; Fig. 9), because the gradient of \( A_{n,j} \) changes with depth and thus affects the \( g_{m,leaf} \) measured. However, if the gradient in \( g_{m,j} \) matches the gradient in \( A_{n,j} \) (null model with \( g_{m,j} \)
gradient; Fig. 9) or $\alpha_j$ compensates so that $A_{n,j}$ is constant across the leaf (null model with $\alpha_j$

... gradient; Fig. 9) then $g_{m,\text{leaf}}$ will respond less to PPFD. Leaves that saturate photosynthesis under

... low light, i.e. with low photosynthetic capacities (low $J_{\text{max},j}$ model; Fig. 9) will have a more even
distribution of $A_{n,j}$ through the leaf and thus less $g_{m,\text{leaf}}$ response to PPFD. In leaves where

... multiple photosynthetic parameters vary across the leaf, more diverse responses of $g_{m,\text{leaf}}$ to

... PPFD occur, including steeper positive responses, or even a slight negative responses where

... $g_{m,\text{leaf}}$ decreases at high PPFD. Furthermore, environments with illumination from both sides of

... the leaf, or with significant diffuse light should lead to a very flat response of $g_{m,\text{leaf}}$ to PPFD.

Finally, significant gradients in $C_i$ across the leaf intercellular airspace would amplify the

... observed $g_{m,\text{leaf}}$ response to PPFD as both $C_i$ and $C_c$ used in the calculation of $g_{m,\text{leaf}}$ would be

... poorly represented by an average value, a point already raised by Parkhurst (1994).

Hence, maximum $g_{m,\text{leaf}}$ can be measured only when all layers are saturated with PPFD,

... implying that illumination from both sides of the leaf is important for leaves with thick cross-

... sections or lots of chlorophyll, or for leaves that lack saturation of $A_{n,\text{leaf}}$ under full sunlight. But

... for many leaves, $g_{m,\text{leaf}}$ measured at full sunlight will be adequate to approximately match

... maximum $g_{m,\text{leaf}}$.

FIGURE 9 COULD BE PLACED HERE.

Implications for photosynthetic limitation analysis, spectral quality responses,

... and canopy photosynthesis

In the model presented here, the layer based value of mesophyll conductance, $g_{m,j}$, does not vary,

... but the leaf value apparently does. Specifically, if $g_{m,j}$ is constant with PPFD, then on a layer

... basis photosynthesis ($A_{n,j}$) is not limited dynamically by $g_{m,j}$ at different PPFD’s. If this effect is
generally true, variation in $g_{m,leaf}$ with changing PPFD does not represent a dynamic limitation to photosynthesis. Thus, photosynthetic limitation analyses, e.g. Grassi & Magnani (2005), should avoid making conclusions based upon data sets in which $g_{m,leaf}$ was measured at varying PPFD. Measurements of daily time courses under natural light conditions should be avoided, as the PPFD effect on $g_{m,leaf}$ might add another confounding effect on the $g_m$ response to the environment. Measuring leaves under natural conditions but above a certain PPFD (e.g. Grassi et al. 2009) may provide a suitable alternative as it is under high PPFD that variations in $g_{m,j}$ and $J_{max,j}$ would lead to least error in the estimates of $g_{m,leaf}$ (Fig. 9), particularly for thin leaves. Limitation analysis seems only appropriate for constant light settings, and most appropriate under high light conditions, especially when comparing leaves of different anatomy (e.g. sun versus shade leaves, developmental stages). Similarly, modelling a variable $g_{m,leaf}$ response to the environment phenomenologically (Yin et al. 2009) would result in apparent limitations to photosynthesis under low light conditions, that are actually RuBP-regeneration limitations and not diffusional limitations.

The theoretical basis for the $g_{m,leaf}$ to PPFD response suggested here could provide an alternative explanation as to why $g_m$ was previously found to respond to blue light (Loreto et al. 2009). The reported response of $g_{m,leaf}$ to blue light was rapid and unrelated to chloroplast movement (Loreto et al. 2009). A possible explanation for the effect may be that the red and blue light applied differed in depth of penetration at the low PPFD used in former experiment (300 µmol m$^{-2}$ s$^{-1}$). If red and blue light penetrated to different depths, as shown previously (e.g. Brodersen & Vogelmann 2010, Evans and Vogelmann 2006 (see also Fig. 7), Oguchi et al. 2011, Vogelmann & Han 2000), then varying $g_{m,j}$ and photosynthetic capacity with depth in the species
used could account for the apparent changes in leaf $g_{m,\text{leaf}}$ reported by Loreto et al. (2009) for blue light.

An analogous model of layer based photosynthesis, to the one suggested here, could be created for layers of leaves in canopies. If so, the emergent response of canopy $g_m$ to light or other environmental factors would also be apparently responsive if the relative photosynthetic contribution of each layer of leaves varied with the environmental factor. This effect would provide alternative explanations to previously observed data. Possible examples are that bulk canopy $g_m$ varied with ABA application (Schäeufele et al. 2011), and could be explained if the photosynthesis varied between leaves (i.e. though an ABA effect), but the $g_{m,\text{leaf}}$ remained constant for leaves of varying age. The large apparent variation in canopy $g_m$ estimated from eddy-covariance (Keenan et al. 2010) seems likely to be highly dependent upon this effect; where differing contributions of leaf layers to eddy-covariance with changing light and drought are highly likely. In that case, the observed limitation on canopy photosynthesis is likely due to RuBP-regeneration (light), and only apparently due to the calculated decreasing canopy $g_m$.

Concluding remarks

Fluxes such as net photosynthetic rate are comprised of the additive contributions of all cells in the leaf. As a result, the true net photosynthesis for a leaf can be unambiguously measured. Mesophyll conductance to CO$_2$ is a conductance, not a flux, having a photosynthetic rate (flux), a potentially variable source ($C_i$ or $C_{ij}$) and a sink ($C_{cj}$) that varies cell by cell (e.g. eqn. 9). For instance, leaf $C_c$ does not exist as a discrete value in any leaf, as each layer must have different values for $C_{cj}$ as pointed out by Parkhurst (1994), but the impact of this on photosynthetic
modelling remains to be fully investigated. In general, this means that $g_{m,\text{leaf}}$ is an emergent property of many leaf anatomical traits adding up to structure in its true three-dimensional nature. Thus, $g_{m,\text{leaf}}$ can appear to respond to environmental variables such as light, despite no structural changes in the basis of CO$_2$ diffusion in the leaf. To our knowledge, the data presented here are the first non-anatomical evidence that indicate that there are gradients in cellular mesophyll conductance across the leaf profile. While the observed responses of $g_{m,\text{leaf}}$ to light are fully consistent with an anatomical, structural or 3D nature of leaf, the results do not preclude a dynamic, regulated response of $g_{m,\text{leaf}}$ to PPFD. If the latter responses exist, then the relative weight of the structural and dynamic responses would vary, the former being fixed in time during leaf growth, while the latter, if present, would allow for a shorter timescale control of CO$_2$ diffusion.

CONFLICTS OF INTEREST

The authors have no conflicts of interest to declare.

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Summary of Lloyd et al.’s (1992) derivation of $g_m$ accounting for structural variations in photosynthetic characteristics across a leaf

Lloyd et al. (1992) presented a mathematically sophisticated approach to calculate $g_{m,leaf}$ to account for variations in photosynthetic characteristics across a leaf, building on what Terashima and Inoue (1985) presented before. Briefly, they considered volume-based variables so that (eqn. A1.1 in the original paper):

$$A_{n,leaf} = \int_0^t A_v dx$$  \hspace{1cm} (eqn. 17)

where $t$ is the thickness of the leaf and $x$ is the distance above the abaxial surface. Keeping this volume-based approach and using carbon isotopic discrimination equations (eqn. A1.2-1.4 in the original paper), they derived an estimated leaf conductance to CO$_2$ as (their eqn. A1.5):

$$g_{m,leaf} = \frac{\left(\int_0^t A_v dx\right)^2}{\int_0^t \frac{A_v^2}{g_m} dx}$$  \hspace{1cm} (eqn. 18)

which can be converted to a leaf area based expression, as the sum of each layer $j$ for a leaf consisting of $l$ layers:

$$g_{m,leaf} = \frac{\left(\sum_{j=1}^l A_{n,j}\right)^2}{\sum_{j=1}^l A_{n,j} / g_{m,j}}$$  \hspace{1cm} (eqn. 19)
This equation of Lloyd et al. (1992) can be simplified to result in eqn. 9. Rearranging eqn. 19 and substituting eqn. 5 for $g_{m,j}$:

$$g_{m,\text{leaf}} = \frac{\left(\sum_{j=1}^{l} A_{n,j}\right)^2}{\sum_{j=1}^{l} A_{n,j}(C_{i,j} - C_{c,j})} \quad \text{(eqn. 20)}$$

If $C_{i,j}$ is assumed to be the same for all layers (i.e. $C_{i}$), then the equation can be rearranged:

$$g_{m,\text{leaf}} = \frac{\left(\sum_{j=1}^{l} A_{n,j}\right)^2}{c_{i}\sum_{j=1}^{l} A_{n,j} - \sum_{j=1}^{l} A_{n,j}c_{c,j}} \quad \text{(eqn. 21)}$$

and dividing the numerator and denominator by $\sum_{j=1}^{l} A_{n,j}$, then:

$$g_{m,\text{leaf}} = \frac{\sum_{j=1}^{l} A_{n,j}}{\frac{c_{i}\sum_{j=1}^{l} A_{n,j} - \sum_{j=1}^{l} A_{n,j}c_{c,j}}{\sum_{j=1}^{l} A_{n,j}}} \quad \text{(eqn. 22)}$$

By simplification, and replacing the numerator with $A_{n,\text{leaf}} (= \sum_{j=1}^{l} A_{n,j})$ we get eqn. 9.

**Alternative derivation for leaf $g_m$ based upon different photosynthetic layers**

A simpler derivation of eqn. 9 starts with defining leaf apparent $g_m$ as:

$$g_{m,\text{leaf}} = \frac{A_{n,\text{leaf}}}{C_{i} - C_{c,\text{wt}}} \quad \text{(eqn. 23)}$$

assuming that $C_{i}$ is the same for all leaf layers. $C_{c,\text{wt}}$ is then the apparent value, the weighted average for the many cell layers. If weighted by layer photosynthesis, similar to eqn. A1.20 (Lloyd et al. 1992):
\[ C_{c, wt} = \frac{\sum_{j=1}^{l} A_{n,j} C_{c,j}}{\sum_{j=1}^{l} A_{n,j}} \]  

(eqn. 24)

As \( A_{n,leaf} \) is a sum of layer \( A_{n,j} \):

\[ A_{n,leaf} = \sum_{j=1}^{l} A_{n,j} \]  

(eqn. 25)

Then substituting eqn. 25 into eqn. 23 we get eqn. 9:

\[ g_{m, leaf} = \frac{A_{n,leaf}}{\sum_{j=1}^{l} A_{n,j} C_{c,j}} \frac{C_{c,j}}{A_{n,leaf}} \]  

(eqn. 9)

This equation has the expected property that when a particular \( A_{n,j} \) tends towards zero, then that layer’s \( C_{c,j} \) is a decreasing component of the calculation of leaf weighted \( C_c \).
**Table 1. Mathematical terms used in the model**

<table>
<thead>
<tr>
<th>Name</th>
<th>Symbol</th>
<th>Value and units</th>
</tr>
</thead>
<tbody>
<tr>
<td>Absorptance of layer $j$</td>
<td>$\alpha_j$</td>
<td>[-]</td>
</tr>
<tr>
<td>Partitioning factor of $\alpha_j$ to PSII</td>
<td>$\beta_j$</td>
<td>0.5 [-]</td>
</tr>
<tr>
<td>Compensation point of layer $j$</td>
<td>$\Gamma^*_{j}$</td>
<td>37.4$^a$ $\mu$mol $\text{mol}^{-1}$</td>
</tr>
<tr>
<td>Light response curvature factor for layer $j$</td>
<td>$\theta_j$</td>
<td>[-]</td>
</tr>
<tr>
<td>Net photosynthesis of leaf and layer $j$</td>
<td>$A_{n,\text{leaf}}, A_{n,j}$</td>
<td>$\mu$mol $m^{-2}$ $s^{-1}$</td>
</tr>
<tr>
<td>Net photosynthesis on volume basis</td>
<td>$A_v$</td>
<td>$\mu$mol $\mu$m$^{-3}$ $s^{-1}$</td>
</tr>
<tr>
<td>Chloroplastic [CO$_2$] of leaf and layer $j$</td>
<td>$C_c,C_{c,j}$</td>
<td>$\mu$mol $\text{mol}^{-1}$</td>
</tr>
<tr>
<td>$C_c$ value for leaf weighted by layer</td>
<td>$C_{c,\text{wt}}$</td>
<td>$\mu$mol $\text{mol}^{-1}$</td>
</tr>
<tr>
<td>Intercellular CO$_2$ concentration of leaf</td>
<td>$C_i$</td>
<td>$\mu$mol $\text{mol}^{-1}$</td>
</tr>
<tr>
<td>A correction for spectral quality of light</td>
<td>$f_j$</td>
<td>0.15 [-]</td>
</tr>
<tr>
<td>Mesophyll conductance of leaf or cell layer $j$</td>
<td>$g_{m,\text{leaf}}, g_{m,j}$</td>
<td>$\text{mol} m^{-2}$ $s^{-1}$</td>
</tr>
<tr>
<td>Irradiance used by PSII of layer $j$</td>
<td>$I_{2,j}$</td>
<td>$\mu$mol $m^{-2}$ $s^{-1}$</td>
</tr>
<tr>
<td>Index representing layers of cells</td>
<td>$j$</td>
<td></td>
</tr>
<tr>
<td>Electron transport rate of layer $j$</td>
<td>$J_j$</td>
<td>$\mu$mol $m^{-2}$ $s^{-1}$</td>
</tr>
<tr>
<td>Maximum electron transport rate of layer $j$</td>
<td>$J_{\text{max},j}$</td>
<td>$\mu$mol $m^{-2}$ $s^{-1}$</td>
</tr>
<tr>
<td>Michaelis constant for Rubisco carboxylation</td>
<td>$K_{c,j}$</td>
<td>272.4$^a$ $\mu$mol $\text{mol}^{-1}$</td>
</tr>
<tr>
<td>Michaelis constant for Rubisco oxygenation</td>
<td>$K_{o,j}$</td>
<td>165.8$^a$ $\text{mol mol}^{-1}$</td>
</tr>
<tr>
<td>Total layers of cells modelled in leaf</td>
<td>$l$</td>
<td>2 to 3</td>
</tr>
<tr>
<td>Oxygen concentration</td>
<td>$O$</td>
<td>200 $\text{mmol mol}^{-1}$</td>
</tr>
<tr>
<td>Photosynthetic photon flux density</td>
<td>PPFD</td>
<td>$\mu$mol $m^{-2}$ $s^{-1}$</td>
</tr>
<tr>
<td>Parameter</td>
<td>Symbol</td>
<td>Unit</td>
</tr>
<tr>
<td>-----------------------------------------------------</td>
<td>----------------</td>
<td>----------------</td>
</tr>
<tr>
<td>PPFD incident on layer $j$</td>
<td>$\text{PPFD}_j$</td>
<td>$\mu\text{mol m}^{-2} \text{s}^{-1}$</td>
</tr>
<tr>
<td>Day respiration of layer $j$</td>
<td>$R_{d,j}$</td>
<td>$\mu\text{mol m}^{-2} \text{s}^{-1}$</td>
</tr>
<tr>
<td>Rate of Rubisco carboxylation of layer $j$</td>
<td>$V_{c,Rubisco,j}$</td>
<td>$\mu\text{mol m}^{-2} \text{s}^{-1}$</td>
</tr>
<tr>
<td>Rate of RuBP-regeneration of layer $j$</td>
<td>$V_{c,RuBP,j}$</td>
<td>$\mu\text{mol m}^{-2} \text{s}^{-1}$</td>
</tr>
</tbody>
</table>

856 a photosynthetic parameters taken from Bernacchi et al. (2002)

857
Table 2. Anatomical measurements of three evenly spaced layers in leaf profiles of *Arbutus* and *Triticum*.

<table>
<thead>
<tr>
<th>Species</th>
<th>Arbutus</th>
<th>Triticum</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Layer</strong></td>
<td><strong>Whole</strong></td>
<td><strong>Adaxial</strong></td>
</tr>
<tr>
<td><strong>Mesophyll thickness (µm)</strong></td>
<td>236 (8)&lt;sup&gt;a&lt;/sup&gt;</td>
<td>148 (18)</td>
</tr>
<tr>
<td><strong>Porosity (%)</strong></td>
<td>25 (2)</td>
<td>11 (1)</td>
</tr>
<tr>
<td><strong>$S_m$ (µm$^2$ µm$^{-2}$)</strong></td>
<td>25.9 (2.4)</td>
<td>9.2 (1.0)</td>
</tr>
<tr>
<td><strong>$S_c$ (µm$^2$ µm$^{-2}$)</strong></td>
<td>21.8 (1.7)</td>
<td>8.2 (0.9)</td>
</tr>
<tr>
<td><strong>$S_c$ / whole leaf $S_c$</strong></td>
<td>0.38</td>
<td>0.38</td>
</tr>
<tr>
<td><strong>$r$</strong>&lt;sup&gt;b&lt;/sup&gt;</td>
<td>1.50</td>
<td>1.42</td>
</tr>
<tr>
<td><strong>Nb. of cell layers</strong></td>
<td>7</td>
<td>1</td>
</tr>
<tr>
<td><strong>$T_{cw}$ (µm)</strong></td>
<td>0.35 (0.04)</td>
<td>0.38 (0.12)</td>
</tr>
<tr>
<td><strong>$T_{cyt}$ (µm)</strong></td>
<td>0.39 (0.03)</td>
<td>0.25 (0.11)</td>
</tr>
<tr>
<td><strong>$g'_{liq}$ (mmol m$^{-2}$ chloroplast s$^{-1}$)</strong></td>
<td>52 (5)</td>
<td>20 (1)</td>
</tr>
<tr>
<td><strong>$g_m$ ($g'_{liq}$ × $S_c$ mol m$^{-2}$ s$^{-1}$)</strong></td>
<td>0.39 (0.04)</td>
<td>0.16 (0.01)</td>
</tr>
</tbody>
</table>

<sup>a</sup> Standard deviation in parenthesis. Five cross-sections measured for each species (two different leaves per species).

<sup>b</sup> Curvature correction factor, based on the method of Thain (1983). The value for *Triticum* is in agreement with values for other grasses e.g. 1.42 for *Oryza* sp. (Giuliani et al. 2013).

<sup>c</sup> Estimated anatomical features made using methods in Evans et al. (2009).
FIGURE LEGENDS

Figure 1. Simulated bulk leaf mesophyll conductance ($g_{m,leaf}$) response to PPFD based upon modelling photosynthesis of multiple layers ($j$) of cells with differential penetration of PPFD with depth (top right). The estimation of $g_{m,leaf}$ is undefined when $A_{n,leaf}$ is close to zero (see eqn. 9), and such values were removed from the top right curve. Detailed simulation values across a three-layer leaf cross section for 65, 100, and 1500 µmol m$^{-2}$ s$^{-1}$ PPFD are shown on the bottom row, with the gradient in light within the leaf represented by the shade gradient from adaxial to abaxial epidermis (general mesophyll cell model of photosynthesis for the path from $C_i$ to $A_{n,j}$ shown on top left). The simulation was run for the null model of a three layer leaf, each with identical photosynthetic parameters, including: $\alpha_j = 0.6$, $g_{m,j} = 0.1$ mol m$^{-2}$ s$^{-1}$, $J_{max,j} = 200$ µmol m$^{-2}$ s$^{-1}$, $\theta_j = 0.6$, and $R_{d,j} = 1$ µmol m$^{-2}$ s$^{-1}$ per layer. The un-italicized value for $g_{m,j}$ was the constant for each layer as set in the model, the italicized value was the weighted value used to calculate total leaf $g_{m,leaf}$ using eqn. 9. Units in the figure: $C_c$ and $C_i$, µmol mol$^{-1}$; $g_m$, mol m$^{-2}$ s$^{-1}$; $A_n$, µmol m$^{-2}$ s$^{-1}$.

Figure 2. Measured responses of leaf total net photosynthesis ($A_{n,leaf}$) and leaf total mesophyll conductance ($g_{m,leaf}$) to PPFD for the two species, with adaxial (black points) or abaxial illumination (gray points). Shaded regions represent the range of fitted values, with the median shown as a solid line, from all the solutions within 10% of the SSE of the best solution, using a three layer model of leaf photosynthesis (fitted values of $\alpha_j$, $g_{m,j}$, $J_{max,j}$, $\theta_j$ and $R_{d,j}$ are shown in Fig. 6). Error bars represent standard errors of the mean for five plants.
Figure 3. Independent validation of the variable J method of measuring bulk leaf mesophyll conductance ($g_m$ variable J) by comparison to simultaneous measurement of $g_m$ using the stable isotope method ($g_m$ SI). Light response curves of $g_m$ for adaxial (black) and abaxial (gray) illumination of leaves are shown for one representative leaf of *Arbutus* (circles) and *Triticum* (squares).

Figure 4. Micrographs of leaf anatomy of *Arbutus* and *Triticum*. Cross sections (top row) are shown at the same scale (width of each image: 341 µm). Dotted lines show where the layers were cut for the parameters measured and presented in Table 2. TEM micrographs (bottom row) present cells from the adaxial (Ad), middle (Mid), and abaxial (Ab) layers for each species (left to right), showing differences in cell wall thickness. Vacuoles in *Arbutus* that are dark grey contain polyphenols as these stain with methylene blue.

Figure 5. The modelled response of net photosynthesis per leaf layer ($A_{n,j}$) for a three layer leaf for adaxial and abaxial illumination for the adaxial (Ad; red), middle (Mid; green), and abaxial (Ab; blue) mesophyll layers. All predicted layer specific light response curves from the solutions within 10% of the SSE of the best solution are presented, and the thick tinted lines represent the median value. Mesophyll conductance values per layer were modelled as constant, but total leaf values varied according to eqn. 9. The modelled curves in these panels result in the fit to the data shown in Fig. 2. Adaxial and middle layers values for *Arbutus* under adaxial illumination are mostly similar and the lines are superimposed.
Figure 6. Distribution of the layer-specific parameters from the optimization solutions that were within 10% of the SSE of the best solution from 500 optimizations using different starting sets of values. Black dots represent the median of layer-specific values, and gray lines show the predicted relative profile based on $S_c$ (for $J_{\text{max},j}$ and $R_{d,j}$) and $g_m$ estimated from leaf anatomy (Table 2). Number of solutions within 10% of the best solution: Arbutus, n = 340; Triticum, n = 33.

Figure 7. Predicted profiles of the fraction of light absorbed for adaxial and abaxial illumination for Arbutus and Triticum. Predicted data (black points) are compared to values for a bifacial (spinach for Arbutus; Evans and Vogelmann 2003) and isobilateral leaf (eucalyptus for Triticum; Evans and Vogelmann 2006). Both these studies generated profiles from the relative $^{14}$C assimilation profiles for blue (solid line) and green light (dashed line). Predicted profiles were computed from the median layer-specific light absorptance value (Fig. 6), and profiles from literature values were computed by dividing the leaf into three layers and summing up the relative absorptance over each layer for both ad- and abaxial profiles.

Figure 8. Proportional sensitivity analysis of simulated $g_{m,\text{leaf}}$ response to PPFD. The values for five parameters in the lower layer of a two layer leaf were decreased by 50%, and the values shown are the percent change in $g_{m,\text{leaf}}$ following that decrease. PPFD values are shown on a log scale to highlight changes under low and high light intensities. The default model of $g_{m,\text{leaf}}$ response to PPFD had both layers of the leaf with the same photosynthetic parameters ($g_{m,j} = 0.1 \, \mu\text{mol m}^{-2} \, \text{s}^{-1}$, $R_{d,j} = 1 \, \mu\text{mol m}^{-2} \, \text{s}^{-1}$, $\alpha_j = 0.75$, $\theta_j = 0.8$, $J_{\text{max},j} = 50 \, \mu\text{mol m}^{-2} \, \text{s}^{-1}$). The estimation
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