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1 Revised submission to *New Phytologist*

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## 3 **Global variability in leaf respiration in relation to climate, plant** 4 **functional types and leaf traits**

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83 **Summary**

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- 85 • Leaf dark respiration ( $R_{\text{dark}}$ ) is an important yet poorly quantified component of the global  
86 carbon-cycle. Given this, we analysed a new global database of  $R_{\text{dark}}$  and associated leaf traits.
- 87 • Data for 899 species were compiled from 100 sites (arctic-to-tropics). Several woody and non-  
88 woody plant functional types (PFTs) were represented. Mixed-effects models were used to  
89 disentangle sources of variation in  $R_{\text{dark}}$ .
- 90 • Area-based  $R_{\text{dark}}$  at the prevailing average-daily growth temperature ( $T$ ) of each site increased  
91 only two-fold from the arctic-to-tropics, despite a 20°C increase in growing  $T$  (8 to 28°C). By  
92 contrast,  $R_{\text{dark}}$  at a standard  $T$  (25°C;  $R_{\text{dark}}^{25}$ ) was three-fold higher in the arctic than tropics, and  
93 two-fold higher at arid than mesic sites. Species and PFTs at cold sites exhibited higher  $R_{\text{dark}}^{25}$   
94 at a given photosynthetic capacity ( $V_{\text{cmax}}^{25}$ ) or leaf nitrogen concentration ([N]) than species at  
95 warmer sites.  $R_{\text{dark}}^{25}$  values at any given  $V_{\text{cmax}}^{25}$  or [N] were higher in herbs than in woody plants.
- 96 • The results highlight variation in  $R_{\text{dark}}$  among species and across global gradients in  $T$  and aridity.  
97 In addition to their ecological significance, the results provide a framework for improving  
98 representation of  $R_{\text{dark}}$  in terrestrial biosphere models (TBMs) and associated land-surface  
99 components of Earth System Models (ESMs).

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102 **Keywords:** Acclimation, aridity, climate models, leaf nitrogen, plant functional types,  
103 photosynthesis, respiration, temperature

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## 105 Introduction

106

107 A challenge for the development of terrestrial biosphere models (TBMs) and associated land surface  
108 components of Earth System Models (ESMs) is improving representation of carbon exchange  
109 between terrestrial plants and the atmosphere, and incorporating biological variation arising from  
110 diversity in plant functional types (PFTs) and climate (Sitch *et al.*, 2008; Booth *et al.*, 2012; Prentice  
111 & Cowling, 2013; Fisher *et al.*, 2014). Accounting for patterns in leaf respiratory CO<sub>2</sub> release in  
112 darkness ( $R_{\text{dark}}$ ) in TBMs and ESMs is crucial (King *et al.*, 2006; Huntingford *et al.*, 2013; Wythers  
113 *et al.*, 2013), since plant respiration – roughly half of which comes from leaves (Atkin *et al.*, 2007)  
114 - releases approximately 60 Pg C yr<sup>-1</sup> (Prentice *et al.*, 2001; Canadell *et al.*, 2007; IPCC, 2013).  
115 Fractional changes in leaf  $R_{\text{dark}}$  as a consequence of climate change can, therefore, have large impacts  
116 on simulated net C-exchange and C-storage for individual ecosystems (Piao *et al.*, 2010) and, by  
117 influencing the CO<sub>2</sub> concentration of the atmosphere, potentially feedback so as to alter the extent  
118 of future global warming (Cox *et al.*, 2000; Huntingford *et al.*, 2013). There is growing acceptance,  
119 however, that leaf  $R_{\text{dark}}$  is not adequately represented in TBMs and ESMs (Huntingford *et al.*, 2013;  
120 Smith & Dukes, 2013), resulting in substantial uncertainty in future climate predictions (Leuzinger  
121 & Thomas, 2011); consequently, there is a need to improve representation of leaf  $R_{\text{dark}}$  in predictions  
122 of future vegetation-climate interactions for a range of possible fossil fuel burning scenarios (Atkin  
123 *et al.*, 2014). Achieving this requires: (1) an analysis of variation in leaf  $R_{\text{dark}}$  along global climate  
124 gradients and among taxa within ecosystems; and, (2) establishing whether relationships between  
125 leaf  $R_{\text{dark}}$  and associated leaf traits vary predictably among environments and plant functional types  
126 (PFTs) (Wright *et al.*, 2004; Reich *et al.*, 2006; Wright *et al.*, 2006; Atkin *et al.*, 2008). PFTs enable  
127 a balance to be struck between the computational requirements of TBMs to minimize the number of  
128 plant groups and availability of sufficient data to fully characterise functional types, versus the reality  
129 that plant species differ widely in trait values. Most TBMs contain at least five PFTs, with species  
130 being organized on the basis of canopy characteristics such as leaf size and life span, physiology,  
131 leaf mass-to-area ratio, canopy height and phenology (Fisher *et al.*, 2014). Although classifications  
132 that are directly trait-based are emerging (Kattge *et al.*, 2011), PFT classifications are still widely  
133 used in TBMs and land surface components of ESMs. As such, discerning the role of PFTs in  
134 modulating relationships between leaf  $R_{\text{dark}}$  and associated leaf traits will provide critical insights.

135 Although our understanding of global variation in leaf  $R_{\text{dark}}$  remains inadequate, it is known  
136 that in natural ecosystems rates vary markedly within and among species, and among PFTs. Surveys  
137 of leaf  $R_{\text{dark}}$  at a common temperature ( $T$ ) of 25°C ( $R_{\text{dark}}^{25}$ ) allow standardized comparisons of  
138 respiratory capacity (and associated investment in mitochondrial protein) to be made among  
139 contrasting sites and species. In a survey of 20 sites around the world, Wright *et al.* (2006) reported

140 a 16-fold variation in mass-based leaf  $R_{\text{dark}}^{25}$ . Importantly, much of the variation in rates of  $R_{\text{dark}}^{25}$   
141 is present within sites among co-occurring species and PFTs, reflecting strong genetic (as opposed  
142 to environmental) control of respiratory flux, as demonstrated by inter-specific comparisons in  
143 controlled-environments (Reich *et al.*, 1998c; Loveys *et al.*, 2003; Xiang *et al.*, 2013) and field  
144 conditions (Bolstad *et al.*, 2003; Tjoelker *et al.*, 2005; Turnbull *et al.*, 2005; Slot *et al.*, 2013).  
145 Differences in demand for respiratory products (e.g. ATP, reducing equivalents and/or carbon  
146 skeletons) from metabolic processes (such as photosynthesis ( $A$ ), phloem loading, N-assimilation  
147 and protein turnover) underpin genotype variations in leaf  $R_{\text{dark}}$  (Lambers, 1985; Bouma *et al.*, 1994;  
148 Bouma *et al.*, 1995; Noguchi & Yoshida, 2008). Consequently, inter-specific variations in leaf  $R_{\text{dark}}$   
149 often scale with photosynthesis (Gifford, 2003; Wright *et al.*, 2004; Campbell *et al.*, 2007), and leaf  
150 nitrogen ([N]) (Ryan, 1995; Reich *et al.*, 1998b). Importantly,  $R_{\text{dark}} \leftrightarrow [\text{N}]$  relationships differ among  
151 PFTs, with  $R_{\text{dark}}$  at a given [N] being higher in forbs than in woody angiosperms and gymnosperms  
152 (Reich *et al.*, 2008).

153 Any analysis of global patterns of leaf  $R_{\text{dark}}$  must consider the impacts of the environment on  
154 respiratory metabolism; here, the impact of  $T$  on  $R_{\text{dark}}$  is of particular interest. Leaf  $R_{\text{dark}}$  is sensitive  
155 to short-term (scale of minutes) changes in  $T$  (Wager, 1941; Atkin & Tjoelker, 2003; Kruse *et al.*,  
156 2011), with the sensitivity declining as leaf  $T$  increases (Tjoelker *et al.*, 2001). With sustained  
157 changes in the prevailing ambient growth  $T$ , leaf  $R_{\text{dark}}$  often acclimates to the new conditions  
158 (Tjoelker *et al.*, 2009; Ow *et al.*, 2010; Dillaway & Kruger, 2011; Slot *et al.*, 2014a), resulting in  
159 higher rates of  $R_{\text{dark}}^{25}$  in cold-acclimated plants (Larigauderie & Körner, 1995; Atkin & Tjoelker,  
160 2003). Such acclimation can occur as quickly as within one to a few days (Atkin *et al.*, 2000) and  
161 can result in leaf  $R_{\text{dark}}$  measured at the prevailing ambient  $T$  ( $R_{\text{dark}}^{\text{amb}}$ ) being nearly identical (i.e.  
162 near-homeostatic) in thermally contrasting environments (Zaragoza-Castells *et al.*, 2008). Another  
163 factor that can influence leaf  $R_{\text{dark}}$  is drought, with rates declining following the onset of drought  
164 (Flexas *et al.*, 2005; Ayub *et al.*, 2011; Crous *et al.*, 2011). However, the response to drought can  
165 vary, with other studies reporting no change (Gimeno *et al.*, 2010) or an increase in  $R_{\text{dark}}^{25}$  with  
166 increasing drought (Bartoli *et al.*, 2005; Slot *et al.*, 2008; Metcalfe *et al.*, 2010). Thus, while  
167 exposure to hot growth conditions is invariably associated with a decline in  $R_{\text{dark}}^{25}$ , there is as yet no  
168 clear consensus on how differences in water availability across sites impact on  $R_{\text{dark}}^{25}$ .

169 As noted above, an overview of global variations in  $R_{\text{dark}}$  is needed to provide benchmarking  
170 data to constrain and test alternative representations of autotrophic respiratory  $\text{CO}_2$  release in TBMs  
171 and the land surface components of ESMs. The data reported by Wright *et al.* (2006) represent the  
172 largest compilation to date, having compared mass-based rates of leaf  $R_{\text{dark}}$  in 208 woody and 60  
173 herb/grass species from 20 contrasting sites, mostly in temperate regions. However, no data were  
174 available for plants growing in upland tropical or arctic ecosystems. Nevertheless, several

175 interesting phenomena were identified, including that rates of  $R_{\text{dark}}^{25}$  (and  $R_{\text{dark}}^{25} \leftrightarrow [\text{N}]$  relationships)  
176 were similar at sites that differ in growth  $T$ ; a similar result was reported in an earlier analysis by  
177 Reich *et al.* (1998b). This observation contrasts with earlier studies that reported higher  $R_{\text{dark}}$  at a  
178 standard measurement  $T$  in plants growing at colder sites (Stocker, 1935; Wager, 1941; Semikhatova  
179 *et al.*, 2007), consistent with thermal acclimation responses of respiratory metabolism (Atkin &  
180 Tjoelker, 2003). A new global database not only requires rates of  $R_{\text{dark}}^{25}$  and  $R_{\text{dark}}^{\text{amb}}$ , but also values  
181 of other leaf traits currently used in TBMs to predict respiration.

182 While there is no single approach to estimating leaf  $R_{\text{dark}}$  in TBMs – Schwalm *et al.* (2010)  
183 reported 15 unique approaches from 21 TBMs – it is common for  $R_{\text{dark}}$  to be related to gross primary  
184 productivity (GPP), either directly as a fraction of GPP, or indirectly as a fraction of maximum  
185 carboxylation capacity, with GPP estimated from enzyme kinetic or stomatal conductance models.  
186 Other models estimate leaf  $R_{\text{dark}}$  from other traits, including  $[\text{N}]$  [e.g. Biome-BGC; Thornton *et al.*  
187 (2002)] and/or vegetation carbon [Lund-Postdam-Jena model (*LPJ*); Sitch *et al.* (2003)]. In the UK  
188 Hadley Centre model *JULES* [Joint UK Land Environment Simulator (Clark *et al.*, 2011)],  $R_{\text{dark}}^{25}$  is  
189 assumed to be proportional to photosynthetic carboxylation capacity at 25°C ( $V_{\text{cmax}}^{25}$ ), with  $V_{\text{cmax}}^{25}$   
190 predicted from PFT-dependent values of leaf  $[\text{N}]$  according to a single  $V_{\text{cmax}}^{25} \leftrightarrow [\text{N}]$  relationship  
191 (Schulze *et al.*, 1994; Cox *et al.*, 1998); *JULES* also provides the opportunity to link terrestrial carbon  
192 cycling to climate prediction. However, as with other models linking  $R_{\text{dark}}^{25}$  to GPP, *JULES* does  
193 not account for climate or PFT-dependent variations in  $R_{\text{dark}}^{25} \leftrightarrow V_{\text{cmax}}^{25} \leftrightarrow [\text{N}]$  relationships. A new  
194 global database will enable assessment of  $R_{\text{dark}}^{25} \leftrightarrow V_{\text{cmax}}^{25} \leftrightarrow [\text{N}]$  (and phosphorous concentrations  
195  $[\text{P}]$ ) relationships, both among PFTs and along climate gradients.

196 Here, using published and unpublished data (Supporting Information, Tables S1 and S2), we  
197 report on a newly compiled global leaf  $R_{\text{dark}}$  and associated traits (*GlobResp*) database. The  
198 *GlobResp* database increases biogeographical and phylogenetic coverage compared to earlier data  
199 sets, and contains information on leaf  $R_{\text{dark}}$  and associated leaf traits for 899 species from 100 sites.  
200 We used the *GlobResp* database to address the following questions. First, do rates of  $R_{\text{dark}}$  at  
201 prevailing ambient  $T$  ( $R_{\text{dark}}^{\text{amb}}$ ) and at a standardized reference  $T$  of 25°C ( $R_{\text{dark}}^{25}$ ) vary with climate  
202 across sites in relation to  $T$  (i.e. thermal environment) and aridity. Second, are the observed patterns  
203 consistent with hypotheses concerning thermal acclimation and adaptation (i.e. evolutionary  
204 response resulting from genetic changes in populations and taxa) of  $R_{\text{dark}}$ . And third, does scaling  
205 between leaf  $R_{\text{dark}}$  and associated leaf traits vary among environments and PFTs? Finally, a key aim  
206 of our study was to predict global variability in  $R_{\text{dark}}^{25}$  from a group of independent input variables,  
207 using data on corresponding leaf traits, climate or a combination of traits and climate; here our aim  
208 was to provide equations that would facilitate improved representation of leaf  $R_{\text{dark}}$  in TBMs and  
209 associated land surface components of ESMs.

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## Materials and Methods

### Compilation of a global database

To create a global leaf respiration and associated leaf traits (*GlobResp*) database, we combined data from recent field campaigns (Supporting Information, Table S1) with previously published data (Table S2). Data were obtained from recent publications (Atkin *et al.*, 2013; Slot *et al.*, 2013; Slot *et al.*, 2014b; Weerasinghe *et al.*, 2014) and the *TRY* trait database (Kattge *et al.*, 2011) that included published studies (Mooney *et al.*, 1983; Oberbauer & Strain, 1985; Oberbauer & Strain, 1986; Chazdon & Kaufmann, 1993; Kamaluddin & Grace, 1993; Kloeppel *et al.*, 1993; García-Núñez *et al.*, 1995; Kloeppel & Abrams, 1995; Zotz & Winter, 1996; Grueters, 1998; Miyazawa *et al.*, 1998; Reich *et al.*, 1998b; Bolstad *et al.*, 1999; Craine *et al.*, 1999; Mitchell *et al.*, 1999; Niinemets, 1999; Wright *et al.*, 2001; Meir *et al.*, 2002; Wright & Westoby, 2002; Veneklaas & Poot, 2003; Wright *et al.*, 2004; Tjoelker *et al.*, 2005; Machado & Reich, 2006; Poorter & Bongers, 2006; Wright *et al.*, 2006; Meir *et al.*, 2007; Swaine, 2007; Sendall & Reich, 2013). The combined database contains data from 100 thermally contrasting sites (899 species representing 136 families, and *c.* 1200 species-site combinations) from biomes ranging from 69°N to 43°S and from sea-level to 3450 m asl (Fig. 1a; Tables 1, 2).

A wide range of terrestrial biomes is represented in the new combined *GlobResp* database (Table 1) along with most of the plant functional types (PFTs) categorised in *JULES* - a land surface component of an Earth System Model (ESM) frameworks (Clark *et al.*, 2011); and in *LPJ* - representing a model with a greater diversity of PFTs from the wider TBM community (Sitch *et al.*, 2003)] (Table 2). Users who would like to use *GlobResp* (to be available via the *TRY* trait database) will also be provided with species classified according to other PFT schemes [including the Sheffield DGVM (Woodward *et al.*, 1998)]. Several PFTs, however, remain poorly represented in *GlobResp*: plants that use the C<sub>4</sub> photosynthetic pathway, boreal deciduous needle-leaved trees (BorDcNI) and tropical herbs/grasses (TrpH – which in the database includes a mixture of species that use either C<sub>3</sub> or C<sub>4</sub> pathways of photosynthesis). Lianas are not yet included in PFT classifications of global TBMs, and are also absent from *GlobResp*, although some data are now emerging for a limited number of sites (Slot *et al.*, 2013). The *GlobResp* database was limited to field-collected data from sites where climate data could be attributed. We excluded data from controlled-environment experiments (e.g. growth cabinets and glasshouses), as well as experiments where atmospheric CO<sub>2</sub>, temperature, irradiance, nutrient supply and/or water supply were manipulated. For each site, long-term climate data were obtained from the *WorldClim* climate database for years 1960–1990, at a resolution of 30 arc seconds, or 1 km at the equator (Hijmans *et al.*, 2005). Aridity indices [AI, ratio of mean annual



245 precipitation (MAP) to potential evapo-transpiration (PET), and hence a lower value of AI indicates  
246 more arid conditions] at each site were estimated according to Zomer *et al.* (2008) using the CGIAR-  
247 CSI Global-PET database (<http://www.cgiar-csi.org>).

248 Mean temperature of the warmest quarter (i.e. warmest three-month period per year; TWQ)  
249 and measuring month (MMT - mean temperature of the month when respiration data were recorded)  
250 were used to characterise the growth  $T$  at each site. Records of the actual measuring month, required  
251 to estimate MMT, were only available for half the sites. Consequently, we used TWQ as a measure  
252 of the growth  $T$ , as most temperate and arctic sites were sampled in summer which corresponded with  
253 the warmest quarter. For tropical sites we also used TWQ, although seasonal  $T$  variation is  
254 comparatively low in tropical regions (Archibold, 1995).

255 Data were collected using similar protocols described herein (Supporting Information  
256 Methods S1) and in published works (Table S2). Outer canopy leaves were sampled early-mid  
257 morning, kept in moist, dark conditions, with  $R_{\text{dark}}$  measured using infra-red gas analysers following  
258 a period of dark-adjustment – typically 30-45 mins (Azcón-Bieto *et al.*, 1983; Atkin *et al.*, 1998).  
259 Only data from mature, fully expanded leaves were included; as such,  $R_{\text{dark}}$  did not reflect the  
260 metabolic demands of biosynthesis associated with localized cell division/expansions processes.  
261 Rather, the measured rates of  $R_{\text{dark}}$  likely reflected demands for respiratory products associated with  
262 cellular maintenance, and potentially phloem loading (Amthor, 2000). We note that the daytime  
263 measured rates of  $R_{\text{dark}}$  may have differed from equivalent fluxes at night (when compared at an  
264 equivalent  $T$ ), reflecting the potential for diel differences in substrate availability and the extent of  
265 sucrose loading.

266 The *GlobResp* database contains  $R_{\text{dark}}$  expressed per unit leaf dry mass and per unit leaf area.  
267 Where available, the database includes values of light-saturated photosynthesis ( $A_{\text{sat}}$ ) and associated  
268 values of internal  $\text{CO}_2$  concentration ( $C_i$ ) and stomatal conductance ( $g_s$ ), leaf mass per area ( $M_a$ ), leaf  
269 nitrogen concentration ( $[N]$ ) and leaf phosphorus concentration ( $[P]$ ).

270

## 271 Temperature normalisation of respiration rates

272 Leaf measurement temperatures ( $T$ ) ranged from 6 to 40°C among sites, with most measured between  
273 16 and 33°C ( $T_1$  in Eqn 1). To enable comparisons of leaf  $R_{\text{dark}}$ , we calculated area- and mass-based  
274 rates both for a common temperature (25°C) and at the growth  $T$  at each site (TWQ and MMT) – see  
275 Methods S2 in Supporting Information for further details. To estimate rates of  $R_{\text{dark}}$  ( $R_2$ ) at a given  $T$   
276 ( $T_2$ ), we calculated rates of  $R_{\text{dark}}$  at 25°C ( $R_{\text{dark}}^{25}$ ), TWQ ( $R_{\text{dark}}^{\text{TWQ}}$ ) and MMT ( $R_{\text{dark}}^{\text{MMT}}$ ) using a  
277 temperature-dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) based on a known rate ( $R_1$ ) at experimental  $T$  ( $T_1$ )  
278 using the equation:  
279

280 
$$R_2 = R_1(3.09 - 0.043 \left[ \frac{(T_2 + T_1)}{2} \right])^{\left[ \frac{T_2 - T_1}{10} \right]}$$
 Eqn 1

281 Calculations of  $R_{\text{dark}}$  at the abover temperatures yielded similar rates, irrespective of whether a  $T$ -  
 282 dependent  $Q_{10}$  or fixed  $Q_{10}$  was used (Supporting Information, Fig. S1).

283

#### 284 Calculation of photosynthetic capacity

285 Given our objective to assess relationships between  $R_{\text{dark}}$  and the carboxylation capacity of Rubisco  
 286 ( $V_{\text{cmax}}$ ), we calculated the  $V_{\text{cmax}}$  for  $C_3$  species (i.e. excluding  $C_4$  species) for all observations where  
 287  $A_{\text{sat}}$  and  $C_i$  values were available (Farquhar *et al.*, 1980; Niinemets, 1999; von Caemmerer, 2000);  
 288 this included all of the previously unpublished data (Table S1) and much of the previously published  
 289 data (Table S2).  $V_{\text{cmax}}$  values were calculated according to:

290

291 
$$V_{\text{cmax}} = (A_{\text{sat}} + R_{\text{light}}) \frac{C_i + K_c [1 + O/K_o]}{C_i - \Gamma^*}$$
 Eqn 2

292

293 where  $\Gamma^*$  is the  $\text{CO}_2$ -compensation point in the absence of non-photorespiratory mitochondrial  $\text{CO}_2$   
 294 release (36.9  $\mu\text{bar}$  at 25°C),  $O$  is the partial pressure of oxygen,  $C_i$  is the inter-cellular  $\text{CO}_2$  partial  
 295 pressure,  $R_{\text{light}}$  is the rate of non-photorespiratory mitochondrial  $\text{CO}_2$  release (here assumed to be  
 296 equal to  $R_{\text{dark}}$ ), and  $K_c$  and  $K_o$  are the Michaelis-Menten constants ( $K_m$ ) of Rubisco for  $\text{CO}_2$  and  $\text{O}_2$ ,  
 297 respectively (von Caemmerer *et al.*, 1994). While the assumption that  $R_{\text{light}} = R_{\text{dark}}$  is unlikely to be  
 298 correct in many situations (Hurry *et al.*, 2005; Tcherkez *et al.*, 2012), estimates of  $V_{\text{cmax}}$  are largely  
 299 insensitive to this assumption. We assumed  $K_c$  and  $K_o$  at 25°C to be 404  $\mu\text{bar}$  and 248 mbar,  
 300 respectively (Evans *et al.*, 1994; von Caemmerer *et al.*, 1994) and that  $K_c$  and  $K_o$  at the measurement  
 301  $T$  could be calculated assuming activation energies ( $E_a$ ) of  $K_c$  and  $K_o$  of 59.4 and 36  $\text{kJ mol}^{-1}$ ,  
 302 respectively (Farquhar *et al.*, 1980). Next, we standardised  $V_{\text{cmax}}$  to 25°C ( $V_{\text{cmax}}^{25}$ ) assuming  $E_a =$   
 303 64.8  $\text{kJ mol}^{-1}$  (Badger & Collatz, 1977) according to:

304

305 
$$V_{\text{cmax}}^{25} = \frac{V_{\text{cmax}}}{\exp\left[\frac{(T-25)E_a}{(298+rT)}\right]}$$
 Eqn 3

306

307 where  $T$  is the leaf temperature at which  $A_{\text{sat}}$  was measured/reported (and thus  $V_{\text{cmax}}$  initially  
 308 estimated), and  $r$  is the gas constant (8.314  $\text{JK}^{-1} \text{mol}^{-1}$ ). Estimates were made for  $C_3$  species only,  
 309 since representation of  $C_4$  plants in our database was minimal (Table 2).

310 For data from unpublished field campaigns (Table S1), leaf area and mass values were  
 311 determined as outlined in Supporting Information (Methods S1); for sites where leaf [N] and [P] were  
 312 both reported, analyses were made using Kjeldahl acid digests (Allen, 1974). For sites where only  
 313 [N] was measured, leaf samples were analyzed by mass spectrometry for total N concentration  
 314 (Loveys *et al.*, 2003); see Table S1 for further details. Details of the N and P analysis procedures

315 used for previously published data can be found in the citations listed in Table S2. Collectively, the  
316 *GlobResp* database contains *c.* 1050 species:site mean values of mass- ( $[N]_m$ ) and area-based leaf  
317 nitrogen concentrations ( $[N]_a$ ), and *c.* 735 species:site mean values of  $[P]_m$  and  $[P]_a$ .

318

## 319 Data analysis

320 Prior to analyses, *GlobResp* data were filtered for statistical outliers. Outlying values were identified  
321 as those falling beyond a central tendency band of twice the interquartile range. Three filters were  
322 applied in sequence to each PFT class separately (using *LPJ* groupings to enable separation of  
323 evergreen and deciduous life histories, and because there were broadly similar numbers of  
324 observations within each *LPJ* PFT category compared to that of *JULES*, where the majority of  
325 observations were within the broadleaved tree (BIT) category). Three filters were applied in the  
326 sequence: (1) mass-based respiration at 25°C ( $R_{\text{dark},m}^{25}$ ); (2) area-based respiration at 25°C ( $R_{\text{dark},a}^{25}$ );  
327 and, (3)  $C_i$  (impacting on the calculation of  $V_{\text{cmax}}$ ). Whenever an outlier was identified, the entire  
328 observational row was removed from the *GlobResp* database. Application of the above filters resulted  
329 in removal of *c.* 3% of the rows from the initial database. Where leaf traits followed an approximate  
330 log-normal distribution, such values were  $\log_{10}$ -transformed before screening for outliers and  
331 subsequent analysis. Analyses were then conducted using: (1) trait averages for unique site:species  
332 combinations; and, where noted, (2), individual rows of data.

333 Bivariate regression was used to explore relationships between area- and mass-based  $R_{\text{dark}}$  and  
334 latitude, TWQ (mean temperature of the warmest quarter, calculated using all data), MMT (mean  $T$   
335 of the month when  $R_{\text{dark}}$  was recorded) and/or AI (ratio of MAP to PET). In addition, backwards-  
336 stepwise regression was used to select the best fitting equation from a starting set of input leaf traits,  
337 climate or the combination of traits plus climate variables; parameters were chosen that exhibited  
338 variance inflation factors (VIF) less than 2.0 (i.e. minimal co-linearity);  $F$ -to-remove criterion was  
339 used to identify best-fit parameters. Multiple regression analyses were then conducted to estimate  
340 predictive equations for the chosen variables. The PRESS statistic (predicted residual error sum of  
341 squares) was used to provide a measure of how well each regression model predicted observed  $R_{\text{dark}}$   
342 values. Relative contributions of leaf trait and climate variables to each regression were gauged from  
343 their standardised partial regression coefficients.

344 Standardised major axis (SMA) analysis was used to determine the best-fit lines ( $\alpha = 0.05$ )  
345 for the key relationships involving  $R_{\text{dark}}^{25}$  both on an area- and mass-basis (Falster *et al.*, 2006; Warton  
346 *et al.*, 2006; Warton *et al.*, 2012) We tested for differences among PFT classes (*JULES*) and site-  
347 based temperature bands (5°C TWQ); to facilitate visual comparison of PFTs, we chose to use the  
348 four PFTs within the *JULES* framework, rather than the larger number of PFTs contained in the *LPJ*  
349 model. Using the *JULES* PFTs also provided an opportunity to assess how changes in growth

350 temperature impacted on bivariate relationships within a PFT [broad leaved trees (BIT)] for which  
351 there was a large number of observations and widespread distribution. We used a mixed-effects linear  
352 model to account for variability in  $R_{\text{dark}}^{25}$  on both area- and mass-bases. Given the hierarchical nature  
353 of the database, the linear mixed-effects model combined fixed and random components (Zuur *et al.*,  
354 2009). The available fixed effect variables included: PFT, leaf traits ( $R_{\text{dark}}^{25}$ ,  $V_{\text{cmax}}^{25}$ , leaf mass per  
355 unit area ( $M_a$ ), [N], [P]) and climate variables (TWQ and AI). Models were run using PFT  
356 classifications from *JULES* and *LPJ*.

357 All continuous explanatory leaf variables were centred on their mean values prior to inclusion.  
358 Co-linearity among leaf variables was tested using VIFs. Model specification and validation was  
359 based on the protocols outlined in Zuur *et al.* (2009) and fitted using the *nlme* package (R package  
360 ver. 3.1–105, R Foundation for Statistical Computing, Vienna, Austria, R Development Core Team  
361 2011). Due to the global nature of the database, species, family and site identifiers were treated as  
362 random rather than fixed effects, placing our focus on the variation contained within these terms,  
363 rather than mean values for each phylogeny/site level. Model comparisons and the significance of  
364 fixed-effects terms were assessed using Akaike's information criterion (AIC).

365 Stepwise and associated multiple linear regressions were conducted using Sigmaplot Statistics  
366 v12 (Systat Software Inc., San Jose, CA, USA). All other statistical analyses and modelling were  
367 conducted using the open-sourced statistical environment 'R' (R Development Core Team, 2011).

368

## 369 **Results**

370

### 371 **Comparison of traits among plant functional types**

372 Across the *GlobResp* database, leaf mass per unit projected leaf area ( $M_a$ ) varied 40-fold (from 19 to  
373 780 g m<sup>-2</sup>), [N]<sub>a</sub> varied 70-fold (from 0.13 to 9.13 g m<sup>-2</sup>) and [P]<sub>a</sub> varied 125-fold, from 10 to 1260  
374 mg m<sup>-2</sup>. In four out of the five *JULES* PFTs (i.e. needle-leaved trees, broad-leaved trees, shrubs and  
375 C<sub>3</sub>-herbs/grasses), ranges of each of  $M_a$ , [N]<sub>a</sub> and [P]<sub>a</sub> values were relatively similar (Figs 2 and S2).  
376 C<sub>4</sub> plants were poorly represented (Table 2), and were generally omitted from subsequent analyses.  
377 On average, shrubs and needle-leaved trees exhibited greater leaf mass per unit area ( $M_a$ ) values than  
378 their broad-leaved tree and C<sub>3</sub> herb/grass counterparts. By contrast, [N]<sub>a</sub> values were relatively  
379 similar among the four PFTs (excluding C<sub>4</sub> plants) (Figs 2 and S2). While [P]<sub>a</sub> values were similar  
380 among broad-leaved trees, C<sub>3</sub> herbs/grasses and shrubs, levels were higher in needle-leaved trees.

381 Area- and mass-based  $V_{\text{cmax}}^{25}$  varied markedly within the four PFTs; needle-leaved trees  
382 exhibited a narrower range of  $V_{\text{cmax}}^{25}$  values compared with the others (Fig. 3a,c). Overall, the average  
383 values of  $V_{\text{cmax}}^{25}$  were relatively similar among the four PFTs. By contrast, average rates of  $R_{\text{dark}}^{25}$

384 differed relatively more among PFTs, being highest in C<sub>3</sub> herbs/grasses, both on an area and mass  
385 basis (Fig. 3b,d).  
386

## 387 Relationships between leaf traits and climate

388 To test whether  $R_{\text{dark}}^{25}$  was related to growth temperature or water availability, we plotted  $R_{\text{dark}}^{25}$   
389 against absolute latitude, TWQ and AI (Figs 4a-c and 4g-i). Against latitude (considering northern  
390 and southern hemispheres separately), area-based  $R_{\text{dark}}^{25}$  ( $R_{\text{dark,a}}^{25}$ ) exhibited a significant, positive  
391 relationship (Table 3), being three-fold faster in arctic than at the equator (Fig. 4a), suggesting, as  
392 expected, that factors other than latitude *per se* play the key roles in determining variations in  $R_{\text{dark,m}}^{25}$ .  
393 A similar pattern in the northern (but not southern) hemisphere was observed for mass-based  $R_{\text{dark}}^{25}$   
394 (Fig. 4g; Table 3). Against TWQ, variations in  $R_{\text{dark,a}}^{25}$  and  $R_{\text{dark,m}}^{25}$  followed trends consistent with  
395 the latitudinal patterns, with rates being fastest at the coldest sites (Figs 4b,h). Negative relationships  
396 were found between both area- and mass-based  $R_{\text{dark}}^{25}$  and AI (Figs 4c,i; Table 3) – recalling that AI  
397 is lowest at the driest sites - with  $R_{\text{dark,a}}^{25} \leftrightarrow \text{AI}$  markedly steeper when data from the wet cool  
398 temperate rainforest site in New Zealand were excluded (Supporting Information, Fig. S2).  
399 Collectively, these results suggest that rates of leaf  $R_{\text{dark}}$  at 25°C are lowest at warm/moist sites near  
400 the equator, and fastest at cold/drier sites at high latitudes.

401 We now consider global patterns of leaf  $R_{\text{dark}}$  at the long-term average ambient growth  $T$  at  
402 each site ( $R_{\text{dark}}^{\text{amb}}$ ), with  $R_{\text{dark}}^{\text{amb}}$  estimated using calculations of  $R_{\text{dark}}$  at TWQ ( $R_{\text{dark}}^{\text{TWQ}}$ ) (Figs 4d-f,  
403 j-l). In the northern hemisphere, both area- and mass-based  $R_{\text{dark}}^{\text{TWQ}}$  decreased with increasing  
404 latitude (Figs 4d,j; Table 3). A similar pattern was observed in the southern hemisphere for mass-  
405 based but not area-based  $R_{\text{dark}}^{\text{TWQ}}$  (Fig. 4d). Both  $R_{\text{dark,a}}^{\text{TWQ}}$  and  $R_{\text{dark,m}}^{\text{TWQ}}$  increased with increasing  
406 TWQ (Fig. 4e,k; Table 3), indicating that rates of  $R_{\text{dark}}^{\text{amb}}$  are likely faster at the warmest sites.  
407 Similarly, the negative  $R_{\text{dark}}^{\text{TWQ}} \leftrightarrow \text{AI}$  association was significant (both on an area and mass-basis;  
408 Fig. 4f,l; Table 3). However, exclusion of mass-based data from the unusually wet site in New  
409 Zealand resulted in there being no significant  $R_{\text{dark,m}}^{\text{TWQ}} \leftrightarrow \text{AI}$  association (Fig. S3). Collectively,  
410  $R_{\text{dark}}^{\text{amb}}$  (both on an area and mass-basis) was faster at the hottest sites in the tropics and mid-latitude  
411 regions. These patterns were consistent whether TWQ or MMT were used as estimates of site-  
412 specific ambient growth  $T$  (Fig. S4).

413 A focus of our study was determining the best function to predict area- and mass-based  $R_{\text{dark}}^{25}$   
414 around the globe from a group of independent input variables. Regression analysis (Table 4) shows  
415 that, based solely on leaf traits (i.e. ignoring climate), 17% and 31% of the variance in  $R_{\text{dark,a}}^{25}$  and  
416  $R_{\text{dark,m}}^{25}$ , respectively, was accounted for using regression equations that included leaf [N] and  
417 area:mass metrics (i.e.  $M_a$  or SLA). Adding leaf [P] did little to improve the proportion of variance  
418 in  $R_{\text{dark}}^{25}$  accounted for by regression; however, [P] replaced [N] in the resultant selected equations  
419 (Table 4). By contrast, addition of  $V_{\text{cmax}}^{25}$  to the available range of leaf traits improved the  $r^2$  of the  
420 resultant regressions (i.e. accounting for 22% and 41% of the variance in  $R_{\text{dark,a}}^{25}$  and  $R_{\text{dark,m}}^{25}$ ,  
421 respectively; Table 4). Climate parameters alone (TWQ, PWQ and/or AI) accounted for only 9-17%

422 of variance in  $R_{\text{dark}}$ . However, combining climate with leaf traits accounted for 35% and 50% of the  
423 variance in  $R_{\text{dark,a}}^{25}$  and  $R_{\text{dark,m}}^{25}$ , respectively (Table 4), with  $M_a$ , TWQ,  $V_{\text{cmax}}^{25}$ , rainfall/aridity and  
424 leaf [P] contributing to variance in  $R_{\text{dark}}$ , largely in that order. Replacing [P] with [N] had little effect  
425 on the  $r^2$  of the resultant linear regressions. Thus, analysis using multiple linear regression strongly  
426 suggests that variations in leaf  $R_{\text{dark}}$  are tied to related variations in leaf structure, chemistry, and  
427 photosynthetic capacity, the thermal environment in the period during which  $R_{\text{dark}}$  measurements were  
428 made, and the average water availability.

429

### 430 Relationships among plant functional types

431 For the  $R_{\text{dark,a}}^{25} \leftrightarrow V_{\text{cmax,a}}^{25}$  association, tests for common slopes revealed no significant differences  
432 among the four *JULES* PFTs, nor did the elevations of those common slopes differ, except for C<sub>3</sub>  
433 herbs/grasses, which exhibited faster rates of  $R_{\text{dark,a}}^{25}$  at a given  $V_{\text{cmax,a}}^{25}$  compared with the other  
434 PFTs (Fig. 5a). Among TWQ classes, there were also no significant differences in slopes, but the  
435 elevation (i.e. y-axis intercept) of the relationships differed systematically when considering all PFTs  
436 collectively (Fig. 5b), and broad-leaved trees alone (Fig. 5c). With respect to the effect of TWQ on  
437  $R_{\text{dark,a}}^{25} \leftrightarrow V_{\text{cmax,a}}^{25}$  relationships, the elevation was similar for the three highest TWQ classes (15-20,  
438 20-25 and >25°C), whereas  $R_{\text{dark,a}}^{25}$  at any given  $V_{\text{cmax,a}}^{25}$  was significantly faster at the two lowest  
439 TWQ classes (Fig. 5b; Table S3). A similar pattern emerged when assessing a single widely-  
440 distributed PFT (broad-leaved trees; Fig. 5c). Thus, in addition to area-based rates of  $R_{\text{dark}}^{25}$  at any  
441 given photosynthetic capacity being fastest in C<sub>3</sub> herbs,  $R_{\text{dark,a}}^{25}$  was also faster in plants growing in  
442 cold environments.

443 Analysed on a mass-basis, tests for common slopes among  $R_{\text{dark,m}}^{25} \leftrightarrow V_{\text{cmax,m}}^{25}$  relationships  
444 revealed significant differences among PFTs and TWQ classes. Among PFTs, the slope of the  
445  $R_{\text{dark,m}}^{25} \leftrightarrow V_{\text{cmax,m}}^{25}$  relationship was greatest in C<sub>3</sub> herbs/grasses and smallest in needle-leaved trees  
446 (Fig. 5d; Table S3); thus, variation in mass-based photosynthetic capacity was matched by greater  
447 variation in leaf  $R_{\text{dark,m}}^{25}$  in herbs/grasses than in needled-leaved trees. Although the effect of TWQ  
448 on  $R_{\text{dark,m}}^{25} \leftrightarrow V_{\text{cmax,m}}^{25}$  was not as consistent as for area-based relationships, in general the pattern was  
449 for  $R_{\text{dark,m}}^{25}$  at any given  $V_{\text{cmax,m}}^{25}$  to be fastest in plants growing in the coldest habitats, particularly  
450 when considering species that exhibit rapid metabolic rates (Fig. 5e,f).

451 Figure 6 shows PFT- and TWQ-dependent variation in  $R_{\text{dark}}^{25} \leftrightarrow [\text{N}]$ . Assessed on a leaf-area  
452 basis, tests for common slopes revealed no significant differences among PFTs (Fig. 6a) or TWQ  
453 classes (Fig. 6b). The elevation of the relationships differed such that at any given leaf  $[\text{N}]_a$ , rates of  
454  $R_{\text{dark,a}}^{25}$  were ranked in order of C<sub>3</sub> herbs/grasses > shrubs > broad-leaved trees = needle-leaved trees  
455 (Table S3). Considering all PFTs collectively, rates of  $R_{\text{dark,a}}^{25}$  at any given  $[\text{N}]_a$  were fastest in the

456 coldest-grown plants, with the overall pattern being one of decreasing  $R_{\text{dark},a}^{25}$  with increasing TWQ  
457 (Fig. 6b). Within broadleaved trees, slopes of  $R_{\text{dark},a}^{25} \leftrightarrow [\text{N}]_a$  relationships differed significantly,  
458 being greater at sites with TWQ values of 15-20°C compared with the two remaining warmer TWQ  
459 categories (Table S3). Hence, for broadleaved tree species with high  $[\text{N}]_a$ ,  $R_{\text{dark},a}^{25}$  was faster in cold  
460 habitats than in their warmer counterparts, at least when considering TWQ classes  $>15^\circ\text{C}$ . Analysing  
461  $R_{\text{dark}}^{25} \leftrightarrow [\text{N}]$  on a mass-basis revealed significant slope differences among PFTs (Fig. 6d) and TWQ  
462 classes (Fig. 6e,f). For the latter, the overall pattern was one of increasing  $R_{\text{dark},m}^{25} \leftrightarrow [\text{N}]_m$  slope in  
463 plants growing at the colder sites.

464

### 465 Mixed effects model analyses

466 Fitting linear mixed-effects models confirmed that the assigned *JULES* PFTs accounted (in  
467 conjunction with assigned random effects) for much of the variation in area-based  $R_{\text{dark}}^{25}$  present in  
468 the *GlobResp* database. For example, a ‘null’ model where fixed effects were limited to four PFT  
469 classes (with species, families and sites treated as random effects) explained 48% of variation in the  
470  $R_{\text{dark},a}^{25}$  response (i.e.  $r^2 = 0.48$ ; Table 5a); for an equivalent model that did not include any random  
471 effects, inclusion of the four PFT classes alone as fixed terms explained 27% of the variation in  
472  $R_{\text{dark},a}^{25}$ . Inclusion of additional fixed terms resulted in an increase in the explanatory power of the  
473 ‘best’ predictive model, such that 70% of variation in  $R_{\text{dark},a}^{25}$  was accounted for via inclusion of  $[\text{N}]_a$ ,  
474  $[\text{P}]_a$ ,  $V_{\text{cmax},a}^{25}$  and TWQ (Fig. 7a, Fig. S3-S5). The variance components of the preferred model, as  
475 defined by the random term (Table 5), indicated that while species and family (Fig. S6) only  
476 accounted for *c.* 8% of the unexplained variance (i.e. the response variance not accounted for by the  
477 fixed terms), *c.* 23% was related to site differences (Fig. S7; Table 5a). Importantly, the linear mixed-  
478 effects model confirmed that  $R_{\text{dark},a}^{25}$  decreased with increasing growth  $T$  (TWQ; Table 5). Using  
479 mass-based variables, the assigned PFTs again accounted for much of the variation in  $R_{\text{dark},m}^{25}$  in the  
480 *GlobResp* database (Table 5), with the ‘null’ model explaining 54% of variation in  $R_{\text{dark},m}^{25}$ . Inclusion  
481 of additional leaf-trait (but not climate) fixed terms resulted in 78% of variation in  $R_{\text{dark},m}^{25}$  being  
482 accounted for (Fig. 7b). For both the area- and mass-based mixed-effect models, the ‘best’ predictive  
483 model (as assessed by AIC criterion; Table S4) yielded predictive PFT-specific equations (Table 6).  
484 Supporting Information provides comparison of models using alternative PFT classifications (*JULES*  
485 & *LPJ*; Table S5); these analyses revealed that replacing *JULES* PFTs with those of *LPJ* did not  
486 improve the power of the predictive models, as shown by the lower AIC values for a model that used  
487 *JULES*-PFTs compared to one using *LPJ*-PFTs (Table S5).

488



## 489 Discussion

490

491 Recognising that leaf respiration is not adequately represented in Terrestrial Biome Models and the  
492 land surface component of Earth System Models (Leuzinger & Thomas, 2011; Huntingford *et al.*,  
493 2013) – reflecting the previous lack of data to constrain estimates of leaf  $R_{\text{dark}}$  - and that improving  
494 predictions of future vegetation-climate scenarios requires global variation in leaf  $R_{\text{dark}}$  to be more  
495 thoroughly characterised (Atkin *et al.*, 2014), we compiled and analysed a new, large global database  
496 of leaf  $R_{\text{dark}}$ , climate conditions and associated traits. Our findings revealed systematic variation in  
497 leaf  $R_{\text{dark}}$  in contrasting environments, particularly regarding to site-to-site differences in growth  
498 temperature and, to a lesser extent, aridity. Importantly, analysis of the *GlobResp* database has  
499 yielded a range of equations (suitable for TBMs and land surface components of ESMs) to predict  
500 variations in  $R_{\text{dark}}$  using information on associated traits (particularly photosynthetic capacity, as well  
501 as leaf structure and chemistry) and growth temperature at each site.

502

### 503 Global patterns in leaf respiration: role of environmental gradients

504 Our results suggest, irrespective of whether rates are expressed on an area or mass basis, that the  
505 global pattern is one of increasing rates of leaf  $R_{\text{dark}}$  with site growth  $T$  (Figs. 4 and S4) when moving  
506 from the cold, dry arctic tundra to the warm, moist tropics. Importantly, however, such increases in  
507 leaf  $R_{\text{dark}}$  are far less than expected given the large range of growth temperatures across sites. One  
508 would expect the variation in TWQ across our sites (*c.* 20°C) to be associated with a *c.* four-fold  
509 increase in  $R_{\text{dark}}^{\text{TWQ}}$  (assuming that  $R_{\text{dark}}$  roughly doubles for every instantaneous 10°C rise in  $T$ ) rather  
510 than the observed *c.* two-fold increases (Fig. 4). Underpinning this constrained variation in  $R_{\text{dark}}^{\text{TWQ}}$   
511 are markedly *faster* area- and mass-based rates of leaf  $R_{\text{dark}}$  at 25°C ( $R_{\text{dark}}^{25}$ ) at the coldest sites, and  
512 *slower*  $R_{\text{dark}}^{25}$  at warmer sites near the equator (Figs 4 and S4).

513 Earlier studies of temperature responses were contradictory: some report faster area- and/or  
514 mass-based rates of  $R_{\text{dark}}^{25}$  at cold sites (Stocker, 1935; Wager, 1941; Semikhatova *et al.*, 1992;  
515 Semikhatova *et al.*, 2007), whilst others have found similar mass-based rates of  $R_{\text{dark}}^{25}$  and  
516  $R_{\text{dark,m}}^{25} \leftrightarrow [N]_{\text{m}}$  relationships in (woody) plants growing in cold and warm habitats (Reich *et al.*,  
517 1998b; Wright *et al.*, 2006). Our new global database, which includes data from Reich *et al.* (1998b)  
518 and Wright *et al.* (2006), contains numerous, previously unpublished data for tropical forest and arctic  
519 tundra sites (Tables 1 and S1), greatly expanding the thermal range and species coverage. Whilst one  
520 might argue that the faster area- and mass-based  $R_{\text{dark}}^{25}$  in cold habitats (Figs 4 and S4) is a result of  
521 the inclusion of tundra herbs/grasses in the *GlobResp* database, growth  $T$  (i.e. TWQ) remained  
522 important when analysing  $R_{\text{dark}}^{25} \leftrightarrow V_{\text{cmax}}^{25}$  and  $R_{\text{dark}}^{25} \leftrightarrow [N]$  relationships within a single, globally-  
523 distributed PFT (broadleaved trees; Figs 5c and 6c). Moreover, the significant negative

524  $R_{\text{dark},a}^{25} \leftrightarrow \text{TWQ}$  and  $R_{\text{dark},m}^{25} \leftrightarrow \text{TWQ}$  relationships (Fig. 4) were maintained when data were restricted  
525 to broadleaved trees (data not shown), albeit with a diminished slope for  $R_{\text{dark},m}^{25} \leftrightarrow \text{TWQ}$   
526 relationships. So, when analysed at the global level, our key finding is that rates of  $R_{\text{dark}}^{25}$  do differ  
527 between cold and warm sites.

528 Faster  $R_{\text{dark}}^{25}$  in plants growing in cold habitats compared to those in warm habitats could  
529 reflect phenotypic (acclimation) or genotypic differences across gradients in growth  $T$ . The ability  
530 of leaf  $R_{\text{dark}}$  to acclimate to sustained changes in growth  $T$  appears widespread among different PFTs  
531 (Atkin & Tjoelker, 2003; Campbell *et al.*, 2007), although there is some evidence that broad-leaved  
532 trees may have a greater capacity to acclimate than their conifer counterparts (Tjoelker *et al.*, 1999).  
533 Acclimation to low growth  $T$  is linked to reversible adjustments in respiratory metabolism (Atkin &  
534 Tjoelker, 2003). Rapid leaf  $R_{\text{dark}}$  are inherent in a number of species characteristic of cold habitats  
535 (Larigauderie & Körner, 1995; Xiang *et al.*, 2013). Similarly, there is evidence that within species,  
536 genotypes from cold habitats can exhibit inherently faster leaf  $R_{\text{dark}}$  than genotypes from warmer  
537 habitats (Mooney, 1963; Oleksyn *et al.*, 1998). However, the pattern (both among and within species)  
538 is far from consistent (Chapin & Oechel, 1983; Atkin & Day, 1990; Collier, 1996).

539 Another site factor that might influence  $R_{\text{dark}}^{25}$  is site water availability or aridity (Figs 4 and  
540 S3; Tables 4 and 5). In our study, faster leaf  $R_{\text{dark}}^{25}$  occurred at the driest sites; similar findings were  
541 reported by Wright *et al.* (2006). Although literature reviews suggest drought-mediated increases in  
542 leaf  $R_{\text{dark}}$  are rare (Flexas *et al.*, 2005; Atkin & Macherel, 2009), there are reports showing that drought  
543 can indeed increase leaf  $R_{\text{dark}}$  (Slot *et al.*, 2008; Metcalfe *et al.*, 2010) and taxa present at drier sites  
544 may also exhibit drought adaptations. However, given our reliance on calculated values of aridity  
545 that may not reflect water availability/loss at field-relevant scales, we suggest that further work is  
546 needed to confirm the extent to which  $R_{\text{dark}}^{25}$  varies in response to aridity gradients.

547

#### 548 Relationships linking respiration to other leaf traits

549 Including  $V_{\text{cmax}}^{25}$  as an explanatory variable markedly improved predictions of  $R_{\text{dark}}^{25}$ , both on an area  
550 and mass basis.  $V_{\text{cmax}}^{25}$  also accounted for a greater proportion of variation in  $R_{\text{dark}}^{25}$  than did leaf [N]  
551 or [P], highlighting the strong functional interdependency between photosynthetic capacity and  
552  $R_{\text{dark}}^{25}$ . Past studies have reported that variation in  $R_{\text{dark}}$  is tightly coupled to variation in  
553 photosynthesis (Reich *et al.*, 1998b; Loveys *et al.*, 2003; Whitehead *et al.*, 2004), underpinned by  
554 chloroplast-mitochondrion interdependence in the light and dark (Krömer, 1995; Noguchi & Yoshida,  
555 2008), and energy costs associated with phloem loading (Bouma *et al.*, 1995). Thus, the simplifying  
556 assumption by *JULES* and other modelling frameworks (Schwalm *et al.*, 2010; Smith & Dukes, 2013)  
557 that  $R_{\text{dark},a}^{25}$  is proportional to  $V_{\text{cmax},a}^{25}$  (Cox *et al.*, 1998) is robustly supported by our global analysis.  
558 However, even though there was no significant  $R_{\text{dark},a}^{25} \leftrightarrow V_{\text{cmax},a}^{25}$  relationship for C<sub>3</sub> herbs/grasses in

559 Figure 5a, overall this PFT exhibited faster rates of  $R_{\text{dark},a}^{25}$  at a given  $V_{\text{cmax},a}^{25}$  compared to other  
560 PFTs (Fig. 5a), with average  $R_{\text{dark},a}^{25}:V_{\text{cmax},a}^{25}$  ratios being 0.078 for C<sub>3</sub> herbs, 0.045 for shrubs, 0.033  
561 for broad-leaved trees and 0.038 for needle-leaved trees. Moreover, area or mass-based  $R_{\text{dark}}^{25}$  at any  
562 given  $V_{\text{cmax}}^{25}$  differed among thermally contrasting sites, being faster at colder sites (Figs 5b,e; Table  
563 S3). Given these issues, it is crucial that in TBMs and ESMs that link  $R_{\text{dark},a}^{25}$  to  $V_{\text{cmax},a}^{25}$ , account is  
564 taken of PFTs and the impact of site growth  $T$  on the balance between respiratory and photosynthetic  
565 metabolism.

566 Our documentation of new predictive  $R_{\text{dark},a}^{25} \leftrightarrow [\text{N}]_a$  relationships, to account for variation  
567 among PFTs and site growth  $T$  (Fig. 6), provides an opportunity to improve the next generation of  
568 ESMs. We found that leaf  $R_{\text{dark}}^{25}$  at any given leaf  $[\text{N}]$  was faster in C<sub>3</sub> forbs/grasses than in their  
569 shrub and tree counterparts (both on an area and mass basis), supporting the findings of Reich *et al.*  
570 (2008). In C<sub>3</sub> herbs/grasses, faster rates of  $R_{\text{dark}}^{25}$  at any given leaf  $[\text{N}]$  likely reflect greater relative  
571 allocation of leaf N to metabolic processes than to structural or defensive roles (Evans, 1989;  
572 Takashima *et al.*, 2004; Harrison *et al.*, 2009), combined with high demands for respiratory products.  
573 In addition to PFT-dependent changes in  $R_{\text{dark}}^{25} \leftrightarrow [\text{N}]$  relationships, we also found that rates of leaf  
574  $R_{\text{dark}}^{25}$  at any given leaf  $[\text{N}]$  were faster in plants growing at colder sites. This finding held when all  
575 PFTs were considered together, and also within the single, widespread PFT of broadleaved trees.  
576 Faster leaf  $R_{\text{dark}}^{25}$  at a given  $[\text{N}]$  therefore appears to be a general trait associated with leaf  
577 development in cold habitats (Atkin *et al.*, 2008).

578

#### 579 Variability in leaf respiration rates within individual ecosystems

580 A key feature of scatterplots such as in Fig. 4 (which presents species means at each site) was the  
581 substantial variation in species-mean values of  $R_{\text{dark}}$  at any given latitude, or TWQ, or indeed, within  
582 any given site (frequently 5-10 fold). This is in line with the diversity often reported in other leaf  
583 functional traits (chemical, structural and metabolic) within natural ecosystems (Wright *et al.*, 2004;  
584 Fyllas *et al.*, 2009; Asner *et al.*, 2014). Furthermore, the range of variation in species-mean values  
585 of  $R_{\text{dark}}$  was far larger than the two-fold shift in mean  $R_{\text{dark}}$  observed along major geographic  
586 gradients. Our understanding of which of these factors account for the wide range of respiratory  
587 rates exhibited by co-existing species is still rather poor (Atkin *et al.*, 2014). At an ecological level,  
588 the wide range in  $R_{\text{dark}}$  may reflect differences among co-existing species [e.g. position along the  
589 ‘leaf economic spectrum’ (Wright *et al.*, 2004); position within the conceptual ‘competitive-stress  
590 tolerator-ruderal (CSR)’ space (Grime, 1977)].

591

## 592 Formulating equations that predict global variability in leaf respiration

593 One of our objectives was to develop equations that accurately predict mean rates of leaf  $R_{\text{dark}}^{25}$   
594 observed across the globe. Our final, parsimonious mixed-effects models accounted for 70% of the  
595 variation in area-based  $R_{\text{dark}}^{25}$  (Fig. 7a) and 78% of the variation in mass-based  $R_{\text{dark}}^{25}$  (Fig. 7b). Such  
596 models provide equations that enable  $R_{\text{dark}}^{25}$  to be predicted using inputs from fixed terms such as  
597 PFT, growth  $T$  and leaf physiology/chemistry. Here, we discuss the fixed effects of the area- and  
598 mass-based models.

599 For the area-based model, PFT category was the most important explanatory factor [e.g. in a  
600 model with no random effects, the *JULES* PFT classification alone accounted for 27% of the  
601 variability in  $R_{\text{dark,a}}^{25}$ ], followed by  $V_{\text{cmax,a}}^{25}$ ,  $[\text{P}]_{\text{a}}$ , TWQ and  $[\text{N}]_{\text{a}}$  (Table 5a). Moreover, a comparative  
602 model that included random components, and where fixed effects were limited to the PFT classes,  
603 was still able to explain 43% of the variation in  $R_{\text{dark,a}}^{25}$ , suggesting that while these PFTs represent a  
604 simplification of floristic complexity, they nevertheless help account for much of the global variation  
605 in area-based  $R_{\text{dark}}^{25}$ .

606 Interestingly, introducing information on phenological habit (i.e. evergreen vs deciduous) and  
607 biome by replacing the *JULES* PFTs with those of *LPJ* did not improve the quality of the predictive  
608 model (Table S5). This may appear counterintuitive, but could have arisen because the additional  
609 information contained in the *LPJ*-PFT classifications was already captured in the ‘best’ predictive  
610 model’s explanatory variables (i.e.  $M_{\text{a}}$ ,  $[\text{N}]_{\text{a}}$ ,  $[\text{P}]_{\text{a}}$ , and TWQ) shown in Table 5.

611 The final ‘best’ predictive model retained  $V_{\text{cmax,a}}^{25}$ , providing further support for a coupling  
612 of photosynthetic and respiratory metabolism (Krömer, 1995; Hoefnagel *et al.*, 1998; Noguchi &  
613 Yoshida, 2008). In terms of leaf chemistry, inclusion of  $[\text{N}]_{\text{a}}$  reflects the coupling of respiratory and  
614 N metabolism (Tcherkez *et al.*, 2005), and energy demands associated with protein turnover (Penning  
615 de Vries, 1975; Bouma *et al.*, 1994; Zagdanska, 1995). Moreover, as  $[\text{N}]_{\text{a}}$  is important to  $V_{\text{cmax,a}}^{25}$ ,  
616 inclusion of  $V_{\text{cmax,a}}^{25}$  in the model may to some extent obscure the role of  $[\text{N}]_{\text{a}}$  *per se*. The significant  
617 interaction of PFT and  $[\text{N}]_{\text{a}}$  demonstrates (Table 5) that variation in leaf  $[\text{N}]_{\text{a}}$  has greater proportional  
618 effects on  $R_{\text{dark,a}}^{25}$  in some PFTs (e.g.  $\text{C}_3$  herbs/grasses) than in others (e.g. broad-leaved trees), for  
619 the reasons outlined above. Retention of  $[\text{P}]_{\text{a}}$  in the preferred model suggests that latitudinal variation  
620 in foliar  $[\text{P}]$  (Fig. S2) plays an important role in facilitating faster rates of leaf  $R_{\text{dark,a}}^{25}$  at the cold high-  
621 latitude sites (Figs 4, S4) whilst limiting rates at P-deficient sites in some regions of the tropics  
622 (Townsend *et al.*, 2007; Asner *et al.*, 2014). These findings are likely to have particular relevance for  
623 predictions of  $R_{\text{dark,a}}^{25}$  in TBMs that include dynamic representation of N and P cycling (Thornton *et*  
624 *al.*, 2007; Zaehle *et al.*, 2014).

625 While PFT category remained an explanatory factor in the final model for mass-based  $R_{\text{dark}}^{25}$   
626 (Table 5),  $V_{\text{cmax,m}}^{25}$  emerged as the single most important factor accounting for variability in  $R_{\text{dark,m}}^{25}$ .

627 Importantly, all climate variables were excluded from the model, including site growth  $T$  (TWQ).  
628 Does this mean that variation in  $R_{\text{dark},m}^{25}$  is unrelated to site growth  $T$ , as previously suggested (Wright  
629 *et al.*, 2006)? Not necessarily; variation in  $R_{\text{dark}}^{25}$  on both area and mass bases was tightly linked to  
630 variation in site growth  $T$  (TWQ, Fig. 5). The absence of TWQ in the mass-based mixed model likely  
631 arose from the influence of site growth  $T$  on leaf  $[N]_m$ , leaf  $[P]_m$  and  $M_a$ ; all three traits vary in response  
632 to differences in site growth  $T$  (Reich & Oleksyn, 2004; Wright *et al.*, 2004; Poorter *et al.*, 2009).

633 In the preferred models for area- and mass-based  $R_{\text{dark}}^{25}$ , little of the response variance not  
634 accounted for by the fixed terms was related to phylogeny, as represented by ‘family’ (Fig. S8); by  
635 contrast, a substantial component (23-73%) of the response variance not accounted for by the fixed  
636 terms was related to differences among sites (Fig. S9). This suggests that other ‘site’ factors  
637 (including environmental and methodological differences) may have played an important role in  
638 determining variation in  $R_{\text{dark},a}^{25}$ . Soil characteristics may be important, including availability of  
639 nutrients such as calcium, potassium and magnesium (Broadley *et al.*, 2004). In addition, rates of  
640  $R_{\text{dark}}^{25}$  are sensitive to prevailing ambient  $T$  and soil water content in the days preceding measurement  
641 (Gorsuch *et al.*, 2010; Searle *et al.*, 2011). Given this, one would not expect long-term climate  
642 averages to fully capture the actual environment experienced by plants.

643

#### 644 Looking forward: improving representation of leaf respiration in Earth System Models

645 The most direct way of improving representation of leaf respiration in TBMs and the land surface  
646 components of ESMs is to formulate equations that describe patterns in  $R_{\text{dark}}^{25}$  using leaf trait and  
647 climate parameters already incorporated into those models. Our study provides PFT-, leaf-trait- and  
648 climate-based equations, depending on which leaf traits are used in a particular model framework to  
649 predict variation in  $R_{\text{dark}}^{25}$  (e.g. area or mass-based  $[N]$ , or photosynthetic capacity, Tables 5, S4 and  
650 S5). Application of such equations would enable prediction of  $R_{\text{dark}}^{25}$  for biogeographical regions for  
651 which the PFT composition is known. The *GlobResp* database will also assist in the development of  
652 land surface models that use a trait-continuum approach, where bivariate trait associations and trade-  
653 offs are included directly in the models, rather than strictly PFT-categorical approach. For an  
654 overview of the issues relevant to incorporation of trait–climate relationships in TBMs, readers are  
655 directed to recent discussion papers (Scheiter *et al.*, 2013; Verheijen *et al.*, 2013; Higgins *et al.*, 2014).

656 Other challenges to incorporating leaf respiration in ESMs include: (i) establishing models of  
657 diel variations in leaf  $R_{\text{dark}}$  – here, understanding the extent to which our daytime measurements of  
658  $R_{\text{dark}}$  differ from fluxes measured at night will be of interest; (ii) accounting for the appropriate level  
659 of thermal acclimation of leaf  $R_{\text{dark}}^{25}$  to dynamic changes in prevailing growth  $T$  and soil moisture at  
660 all geographical positions; and, (iii) identifying the extent to which light inhibition of leaf respiration  
661 (Kok, 1948; Brooks & Farquhar, 1985; Hurry *et al.*, 2005) varies among PFTs and biomes, over the

662 range of leaf  $T$ s experienced by leaves during the day. Although much progress has been made (King  
663 *et al.*, 2006; Atkin *et al.*, 2008; Smith & Dukes, 2013; Wythers *et al.*, 2013), accounting for  
664 temperature acclimation and light inhibition of leaf  $R$  in TBMs and associated land surface  
665 components of ESMs remains a considerable challenge (Atkin *et al.*, 2014). The equations we  
666 provide here that predict current biogeographical variations in leaf  $R_{\text{dark}}$  at a standard  $T$  (typically  
667 25°C) are driven by some unquantified combination of acclimation responses and genotypic  
668 (adaptive) differences. Further work is needed, however, to establish criteria that will enable  
669 environment and genotypic variations in light inhibition of leaf respiration to be predicted; here,  
670 recent studies linking light inhibition to photorespiratory metabolism (Griffin & Turnbull, 2013;  
671 Ayub *et al.*, 2014) may provide directions for future research. Achieving these goals will be assisted  
672 by compilation of data not only from the sites shown in Figure 1, but also from geographic regions  
673 currently poorly represented; additional data from Africa, Asia and Europe are needed to enable  
674 global historical biogeographic/phylogenetic effects on leaf  $R_{\text{dark}}$  to be tested. In the long term, a  
675 wider goal is development of a mechanistic model that accounts for genotypic-developmental-  
676 environmentally-mediated variations in leaf  $R_{\text{dark}}$ .

677 Currently, many TBM and ESMs predict photosynthetic capacity ( $V_{\text{cmax}}^{25}$ ) and  $R_{\text{dark}}^{25}$  based  
678 on assumed [N] values for each PFT. In using this approach, differences among plants within a PFT  
679 (e.g. genotypic differences and/or plasticity responses to the growth environment) are unspecified.  
680 Our mixed-effects models suggest that PFTs capture a substantial amount of species variation across  
681 diverse sites and their use is reasonable as a first approximation for the purposes of modelling. In the  
682 application of PFT-based modelling, the growth  $T$ -dependent (TWQ) variations in  $R_{\text{dark}}^{25}$  within  
683 widely distributed PFTs (e.g. broadleaved trees) provide a means to predict  $T$ -adjustments in  $R_{\text{dark}}$  at  
684 the global scale. For example, predicted  $R_{\text{dark}}^{25}$  declines 18% from 1.0 to 0.82  $\mu\text{mol m}^{-2} \text{s}^{-1}$  when site  
685 temperature (TWQ) increases from 20 to 25°C (Table 6). Assuming a static PFT (e.g. no species  
686 turnover or differential acclimation/adaptation), these new equations (Table 6, and associated ESM  
687 equations in Table S4) provide a first-order approximation of the acclimation response of  $R_{\text{dark}}^{25}$  of a  
688 given PFT to a cooler past world, or warmer future world. They also demonstrate that predictions  
689 based on PFT, leaf traits and TWQ provide a powerful improvement in the representation of leaf  
690 respiration in ESMs that seek to describe the role of terrestrial ecosystems in an evolving global  
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692

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710

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## Supporting Information

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1126 Additional supporting information may be found in the online version of this article.

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1129 Methods S1: Sampling methods and measurements protocols for previously unpublished data.

1130 Methods S2: Details on methodology used to temperature normalize respiration rates

1131 Table S1. Details on unpublished databases used in global data base of  $R_{\text{dark}}$ .

1132 Table S2. Details on published databases used in global data base of  $R_{\text{dark}}$ .

1133 Table S3. Standardized Major Axis regression slopes for relationships in Figs 5 & 6.

1134 Table S4. Comparison of mixed-effects models with area-based  $R_{\text{dark}}^{25}$  as the response variable

1135 Table S5. Comparison of mixed-effects models using different plant functional types (PFT)  
1136 classifications, with  $R_{\text{dark}}^{25}$  as the response variable.

1137 Figure S1. Comparison of  $R_{\text{dark}}^{25}$ , calculated assuming either a fixed  $Q_{10}$  or a  $T$ -dependent  $Q_{10}$

1138 Figure S2. Relationships between leaf structural/chemical composition traits and TWQ

1139 Figure S3.  $R_{\text{dark}}$ -aridity index relationships, excluding data from a high-rainfall site in NZ

1140 Figure S4.  $R_{\text{dark}}$ -MMT relationships for those sites where the month of measurement was known

1141 Figure S5. Testing key assumptions for mixed effects models –heterogeneity and normality.

1142 Figure S6. Model validation graphs for the area-based mixed effects model

1143 Figure S7. Standardised residuals against fitted values for variables not used in mixed model

1144 Figure S8. Dotchart of the area-based mixed model's random intercepts by Family

1145 Figure S9. Dotchart of the area-based mixed model's random intercepts by site

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**Table 1. Sample sites and climate conditions at which leaf dark respiration ( $R_{dark}$ ) was measured.**

Country/Region	Biomes	Altitude (m asl)	MAT (°C)	TWQ (°C)	MAP (mm)	PWQ (mm)	AI	No. species	PFTs present ( <i>JULES</i> )
USA-AK	Tu	720	-11.3	8.2	225	113	0.61	37	BIT, C3H, S
Russia-Siberia	BF	217	-10.8	15.4	254	122	0.46	3	BIT, NIT
USA-CO	Tu	3,360	-2.6	7.5	811	203	1.20	10	BIT, C3H, NIT, S
USA-MN	BF, TeDF, TeG	365	4.4	18.4	735	303	0.87	53	BIT, C3H, C4H, NIT, S
USA-IA	TeDF	385	7.1	20.2	865	315	0.83	11	BIT, NIT
USA-WI	TeDF, TeG	293	7.7	20.6	880	315	0.93	15	BIT, C3H, NIT
USA-MI	TeDF	200	8.6	19.9	944	268	0.98	1	NIT
Germany	TeDF	60	9.1	17.2	704	190	0.92	9	BIT, NIT
USA-NY	TeDF	225	9.4	20.8	1,173	308	1.20	3	BIT
USA-PA	TeDF	355	9.5	19.9	915	262	0.91	3	BIT
Spain	TeW	1,017	10.7	19.2	487	99	0.48	1	BIT
Australia-TAS	TeRF	144	11.0	14.7	1,325	211	1.58	14	BIT, S
Chile	TeRF	434	11.1	15.4	1,467	103	1.40	18	BIT, NIT
USA-TN	TeDF	775	11.2	20.1	1,554	389	1.34	13	BIT, C3H, NIT, S
New Zealand	TeRF	202	11.3	15.9	4,014	1,011	4.50	16	BIT, NIT, S
USA-NC	TeDF	850	11.4	20.0	1,852	444	1.52	15	BIT, NIT
USA-NM	Sa	1,620	12.5	22.2	275	127	0.19	9	BIT, NIT, S
Australia-ACT	TeW	572	13.0	20.7	722	271	0.58	6	BIT, NIT, S
Japan	TeDF	20	14.9	23.7	1,619	433	1.92	4	BIT
Sth Africa	TeW	600	16.6	21.0	754	67	0.57	5	BIT, S
Peru-Andes	TrRF_up	2,380	16.7	17.7	1,297	373	0.79	82	BIT, C3H
Australia-SA	TeW	35	17.3	23.6	255	52	0.17	10	BIT, C3H, S
Australia-NSW	TeW	140	17.3	23.2	820	215	0.29	70	BIT, C3H, C4H, NIT, S
USA-SC	TeDF	3	17.7	25.8	1,339	469	1.02	10	BIT, C3H, NIT, S
Australia-WA	TeW	204	18.7	24.5	463	47	0.32	55	BIT, C3H, S
Australia-FNQ	TrRF_lw	513	22.4	25.1	1,990	934	1.35	45	BIT, S
Cameroon	TrRF_lw	550	24.0	24.8	1,729	417	1.13	6	BIT, C3H
Venezuela	TrRF_lw	492	24.4	24.7	3,092	693	1.61	10	BIT, S
Bolivia	TrRF_lw	400	25.3	27.0	1,020	436	0.57	50	BIT
Suriname	TrRF_lw	215	25.4	26.3	2,224	165	1.37	25	BIT, C3H, C4H, S
Peru-Amazon	TrRF_lw	164	25.4	26.2	2,567	828	1.50	214	BIT, S
Bangladesh	TrRF_lw	21	25.5	28.5	1,970	736	1.34	1	BIT
Costa Rica	TrRF_lw	135	25.7	26.7	4,141	747	2.64	2	BIT, S
French Guiana	TrRF_lw	21	25.8	26.2	2,824	222	1.88	70	BIT
Malaysia-Borneo	TrRF_lw	20	26.7	27.1	2,471	501	1.64	29	BIT, S
Brazil-Amazon	TrRF_lw	115	27.0	27.6	2,232	401	1.39	9	BIT
Panama	TrRF_lw	98	27.0	27.7	1,822	300	1.19	18	BIT
Niger	Sa	280	28.2	31.4	618	55	0.30	3	BIT, S

1149 Sites shown in order from increasing mean annual temperature (MAT). Where multiple sites were found  
 1150 within a region, values represent the mean values of all sites, weighted for the number of species at each site  
 1151 (see Tables S1 and S2 in Supporting Information for further details). Data on climate are from the *WorldClim*  
 1152 data base (Hijmans *et al.*, 2005). Number of species measured at each site are shown, as are the number of  
 1153 observational rows of data contained in the *GlobResp* database. For the latter, an observational row  
 1154 represents individual measurements for all unpublished data (See Table S1 in Supporting Information), while  
 1155 for published data (SI Table S2) observational rows in many cases represent mean values of species:site  
 1156 combinations. *JULES* (Clark *et al.*, 2011) plant functional types (PFTs) at each site shown, according to: BIT,  
 1157 broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; C4H, C<sub>4</sub> metabolism herb/grass; NIT, needle-leaved tree;  
 1158 S, shrub. Biome classes: BF, boreal forests; TeDF, temperate deciduous forest; TeG, temperate grassland;  
 1159 TeRF, temperate rainforest; TeW, temperate woodland; TrRF\_lw, lowland tropical rainforest (<1500 asl);  
 1160 TrRF\_up, upland tropical rainforest (>1500 asl); Tu, tundra. Abbreviations: mean temperature of the warmest  
 1161 quarter (i.e. warmest 3-month period per year; TWQ), mean annual precipitation (MAP), mean precipitation  
 1162 of the warmest quarter (PWQ), aridity index (AI) calculated as the ratio of MAP to mean annual potential  
 1163 evapotranspiration (UNEP, 1997; Zomer *et al.*, 2008). Australia-ACT, Australian Capital Territory; Australia-  
 1164 FNQ, Far North Queensland; Australia-NSW, New South Wales; Australia-TAS, Tasmania; Australia-WA,  
 1165 Western Australia; USA-AK, Alaska; USA-CO, Colorado; USA-MN, Minnesota; USA-IA, Iowa; USA-WI,  
 1166 Wisconsin; USA-MI, Michigan; USA-PA, Pennsylvania; USA-NY, New York; USA-NC, North Carolina; USA-TN,  
 1167 Tennessee; USA-NM, New Mexico; USA-SC, South Carolina.

**Table 2. Details of plant functional types (PFTs) contained in *GlobResp* database.**

ESM framework	Plant functional types (PFTs)	No. sites	Min. latitude	Max. latitude	No. species
JULES	BIT	94	-43.42	68.63	642
	C3H	14	-34.04	68.63	75
	C4H	3	-33.84	45.41	8
	NIT	20	-43.31	62.25	24
	S	32	-43.42	68.63	124
LPJ	BorDcBl	10	40.05	68.63	18
	BorDcNI	3	33.33	62.25	2
	BorEvNI	6	40.05	62.25	10
	TmpDcBl	25	-43.41	50.60	46
	TmpEvBl	33	-43.42	68.63	193
	TmpEvNI	13	-43.31	50.60	18
	TmpH	12	-34.04	68.63	79
	TrpDcBl	20	-15.78	13.20	50
	TrpEvBl	39	-17.68	24.20	468
TrpH	3	-13.11	3.38	4	

PFTs for two Earth System Model frameworks are shown: *JULES* (Clark *et al.*, 2011) and *LPJ* (Sitch *et al.*, 2003). For each PFT, the number of field sites and species are shown, as is the maximum absolute latitude and longitude of the PFT distribution. For *JULES*, the following PFTs are shown: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; C4H, C<sub>4</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. For *LPJ*, the following PFTs are shown: BorDcBl, boreal deciduous broad-leaved tree/shrub; BorDcNI, boreal deciduous needle-leaved tree/shrub; BorEvNI, boreal evergreen needle-leaved tree/shrub; TmpDcBl, temperate deciduous broad-leaved tree/shrub; TmpEvBl, temperate evergreen broad-leaved tree/shrub; TmpEvNI, temperate evergreen needle-leaved tree/shrub; TmpH, temperate herb/grass; TrpDcBl, tropical deciduous broad-leaved tree/shrub; TrpEvBl, tropical evergreen broad-leaved tree/shrub; TrpH, tropical herb/grass. Note: in some cases, individual species occurred at multiple sites in multiple biomes. Finally, an overwhelming majority of the shrubs (S) were evergreen (123 species:site combinations) compared to deciduous shrubs (11 species:site combinations)

**Table 3. Correlations between log<sub>10</sub> transformed leaf respiration ( $R_{\text{dark}}$ ) and location/climate (see Figure 4).**

Response variable	Latitude								TWQ, both hemispheres								Aridity index, both hemispheres							
	Hemi sphere	df	p value	r <sup>2</sup>	Intercept	Slope	CI slope, lower	CI slope, higher	df	p value	r <sup>2</sup>	Intercept	Slope	CI slope, lower	CI slope, higher	df	p value	r <sup>2</sup>	Intercept	Slope	CI slope, lower	CI slope, higher		
$\log R_{\text{dark,a}}^{25}$	N <sup>th</sup>	404	< 0.0001	0.189	-0.130	0.005	0.004	0.006	1,104	< 0.0001	0.139	0.467	-0.019	-0.022	-0.016	1,069	< 0.0001	0.038	0.119	-0.069	-0.090	-0.048		
	S <sup>th</sup>	698	< 0.0001	0.071	-0.063	0.005	0.004	0.007																
$\log R_{\text{dark,a}}^{\text{TWQ}}$	N <sup>th</sup>	404	< 0.0001	0.039	-0.028	-0.002	-0.003	-0.001	1,104	< 0.0001	0.066	-0.329	0.013	0.010	0.015	1,069	< 0.0001	0.016	0.011	-0.043	-0.063	-0.023		
	S <sup>th</sup>	698	0.740	0.000	N/A	N/A	N/A	N/A																
$\log R_{\text{dark,m}}^{25}$	N <sup>th</sup>	421	< 0.0001	0.148	0.908	0.006	0.004	0.007	1,111	< 0.0001	0.074	1.342	-0.015	-0.019	-0.012	1,076	< 0.0001	0.019	1.065	-0.054	-0.077	-0.031		
	S <sup>th</sup>	688	0.824	0.000	N/A	N/A	N/A	N/A																
$\log R_{\text{dark,m}}^{\text{TWQ}}$	N <sup>th</sup>	421	0.006	0.018	1.008	-0.002	-0.003	-0.001	1,111	< 0.0001	0.083	0.546	0.016	0.013	0.019	1,076	0.031	0.004	0.950	-0.026	-0.049	-0.002		
	S <sup>th</sup>	688	< 0.0001	0.061	0.991	-0.005	-0.007	-0.004																

For each correlation between the y-axis leaf trait and x-axis location/climate parameter, the number of degrees of freedom (df), probability value (*p-value*) and coefficient of determination ( $r^2$ ) and 95% confidence intervals (CI) are shown. Traits shown are:  $R_{\text{dark,a}}^{25}$  and  $R_{\text{dark,a}}^{\text{TWQ}}$ , predicted area-based leaf  $R_{\text{dark}}$  rates ( $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ) at 25°C and TWQ (mean daily temperature of the warmest quarter), respectively; leaf  $R_{\text{dark,m}}^{25}$  and  $R_{\text{dark,m}}^{\text{TWQ}}$ , predicted mass-based  $R_{\text{dark}}$  rates ( $\text{nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$ ) at 25°C and TWQ, respectively. TWQ at each site were obtained using site information and the *WorldClim* data base (Hijmans *et al.*, 2005). Aridity index calculated as the ratio of mean annual precipitation (MAP) to mean annual potential evapotranspiration (PET) (UNEP, 1997; Zomer *et al.*, 2008). Abbreviations: N/A, not applicable.

**Table 4. Regression equations expressing area- and mass-based leaf dark respiration at 25°C ( $R_{\text{dark}}^{25}$ ) as function of other leaf traits and site climate.**

Dependent variable	Input: independent variables (Backwards-Stepwise Regression)	Output: selected equations (Multiple Linear Regression)	Multiple linear regression parameters							
			$n$	$r^2$	PRESS statistic	Standardized partial regression coefficients				
						$\beta_1$	$\beta_2$	$\beta_3$	$\beta_4$	$\beta_5$
	Leaf traits (all $\log_{10}$ ): $[N]_a, M_a$	$\log_{10}R_{\text{dark},a}^{25} = -0.469 + (0.329*\log_{10}N_a) + (0.204*\log_{10}M_a)$	1038	0.168	57.71	0.270 ( $\log_{10}N_a$ )	0.186 ( $\log_{10}M_a$ )			
	Leaf traits (all $\log_{10}$ ): $[N]_a, [P]_a, M_a$	$\log_{10}R_{\text{dark},a}^{25} = 0.076 + (0.304*\log_{10}P_a) + (0.140*\log_{10}M_a)$	730	0.156	40.95	0.338 ( $\log_{10}P_a$ )	0.112 ( $\log_{10}M_a$ )			
Area-based $\log_{10} R_{\text{dark},a}^{25}$	Leaf traits (all $\log_{10}$ ): $[N]_a, [P]_a, M_a, V_{\text{cmax},a}^{25}$	$\log_{10}R_{\text{dark},a}^{25} = -0.241 + (0.235*\log_{10}P_a) + (0.050*\log_{10}M_a) + (0.290*\log_{10}V_{\text{cmax},a}^{25})$	703	0.221	34.79	0.269 ( $\log_{10}P_a$ )	0.041 ( $\log_{10}M_a$ )	0.285 ( $\log_{10}V_{\text{cmax}}$ )		
	Climate parameters: TWQ, PWQ, AI	$\log_{10}R_{\text{dark},a}^{25} = 0.451 - (0.0153*TWQ) - (0.00016*PWQ)$	1114	0.171	61.86	-0.297 (TWQ)	-0.196 (PWQ)			
	Leaf traits (all $\log_{10}$ ) and climate parameters: $[N]_a, [P]_a, M_a, V_{\text{cmax},a}^{25}, TWQ, PWQ$	$\log_{10}R_{\text{dark},a}^{25} = -0.563 + (0.292*\log_{10}M_a) + (0.119*\log_{10}P_a) + (0.221*V_{\text{cmax},a}^{25}) - (0.0147*TWQ) - (0.00012*PWQ)$	703	0.353	29.06	0.238 ( $\log_{10}M_a$ )	0.136 ( $\log_{10}P_a$ )	0.243 ( $\log_{10}V_{\text{cmax}}$ )	-0.304 (TWQ)	-0.165 (PWQ)
	Leaf traits (all $\log_{10}$ ): $[N]_m, SLA$	$\log_{10}R_{\text{dark},m}^{25} = 0.0932 + (0.475*\log_{10}SLA) + (0.364*\log_{10}N_m)$	1037	0.314	57.78	0.392 ( $\log_{10}SLA$ )	0.244 ( $\log_{10}N_m$ )			
	Leaf traits (all $\log_{10}$ ): $[N]_m, [P]_m, SLA$	$\log_{10}R_{\text{dark},m}^{25} = 0.495 + (0.556*\log_{10}SLA) + (0.333*\log_{10}P_m)$	730	0.336	40.68	0.396 ( $\log_{10}SLA$ )	0.315 ( $\log_{10}P_m$ )			
Mass-based $\log_{10} R_{\text{dark},m}^{25}$	Leaf traits (all $\log_{10}$ ): $[N]_m, [P]_m, SLA, V_{\text{cmax},m}^{25}$	$\log_{10}R_{\text{dark},m}^{25} = -0.061 + (0.432*\log_{10}SLA) + (0.264*\log_{10}P_m) + (0.274*\log_{10}V_{\text{cmax},m}^{25})$	703	0.407	34.80	0.307 ( $\log_{10}SLA$ )	0.252 ( $\log_{10}P_m$ )	0.263 ( $\log_{10}V_{\text{cmax}}$ )		
	Climate parameters: TWQ, PWQ, AI	$\log_{10}R_{\text{dark},m}^{25} = 1.353 - (0.0157*TWQ) - (0.000018*AI)$	1121	0.087	83.22	-0.276 (TWQ)	0.112 (AI)			
	Leaf traits (all $\log_{10}$ ) and climate parameters: $[N]_m, [P]_m, SLA, V_{\text{cmax},m}^{25}, TWQ, AI$	$\log_{10}R_{\text{dark},m}^{25} = 0.249 + (0.526*\log_{10}SLA) + (0.0705*\log_{10}P_m) + (0.281*\log_{10}V_{\text{cmax},m}^{25}) - (0.0184*TWQ) - (0.000015*AI)$	703	0.497	29.72	0.374 ( $\log_{10}SLA$ )	0.067 ( $\log_{10}P_m$ )	0.270 ( $\log_{10}V_{\text{cmax}}$ )	-0.333 (TWQ)	0.111 (AI)

**All leaf trait data were  $\log_{10}$  transformed.** To select the best fitting equation from a group of input independent variables (e.g. leaf trait, climate or the combination of trait plus climate), data were explored using *Backwards-Stepwise Regression* – this revealed that chosen parameters exhibited *variance inflation factors* (VIF) less than 2.0 (i.e. minimal multi-collinearity); it also identified best-fit parameters (using *F-to-remove* criterion). Thereafter, multiple regression analyses were conducted to estimate predictive equations for the chosen variables. All selected variables were significant ( $P < 0.001$ ). The *PRESS* statistic (predicted residual error sum of squares) provides a measure of how well each regression model predicts the observations, with smaller *PRESS* indicating better predictive capability. Relative contributions of leaf trait and climate variables to each regression can be gauged from their standardized partial regression coefficients ( $\beta_1$ - $\beta_5$ , depending on model equation). Abbreviations:  $M_a$ , leaf mass per unit leaf area ( $\text{g m}^{-2}$ );  $SLA$ , leaf area per unit leaf mass ( $\text{m}^2 \text{kg}^{-1}$ );  $[N]_a$  and  $[N]_m$ , area- ( $\text{g m}^{-2}$ ) and mass-based ( $\text{mg g}^{-1}$ ) leaf nitrogen concentration, respectively;  $[P]_a$  and  $[P]_m$ , area- ( $\text{g m}^{-2}$ ) and mass-based ( $\text{mg g}^{-1}$ ) leaf phosphorus concentration, respectively;  $V_{\text{cmax},a}^{25}$  ( $\mu\text{mol CO}_2 \text{m}^{-2} \text{s}^{-1}$ ) and  $V_{\text{cmax},m}^{25}$  ( $\text{nmol CO}_2 \text{g}^{-1} \text{s}^{-1}$ ), predicted area- and mass-based capacity for  $\text{CO}_2$  fixation by Rubisco at 25°C, respectively;  $R_{\text{dark},a}^{25}$  ( $\mu\text{mol CO}_2 \text{m}^{-2} \text{s}^{-1}$ ) and  $R_{\text{dark},m}^{25}$  ( $\text{nmol CO}_2 \text{g}^{-1} \text{s}^{-1}$ ), predicted area- and mass-based leaf  $R_{\text{dark}}$  rates at 25°C, respectively; mean temperature of the warmest quarter (i.e. warmest 3-month period per year; TWQ, °C), mean annual precipitation (MAP,  $\text{mm yr}^{-1}$ ), mean precipitation of the warmest quarter (PWQ), aridity index (AI) calculated as the ratio of MAP to mean annual potential evapotranspiration (UNEP, 1997; Zomer *et al.*, 2008). TWQ at each site were obtained using site information and the *WorldClim* data base (Hijmans *et al.*, 2005).

**Table 5. Two linear mixed-effects models ('best' predictive model and a 'null', PFT only model), with (a) area-based ( $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ) and (b) mass-based ( $\text{nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$ ) leaf respiration at  $25^\circ\text{C}$  ( $R_{\text{dark,a}}^{25}$  and  $R_{\text{dark,m}}^{25}$ , respectively) as the response variables, each showing fixed and random effects. See Table 6 for PFT-specific equations are shown that can be used to predict variability in  $R_{\text{dark,a}}^{25}$  and  $R_{\text{dark,m}}^{25}$  based on 'best' models.**

(a) Area-based model							(b) Mass-based model						
Fixed effects							Fixed effects						
'Best' predictive model (PFTs, leaf traits and climate)				'Null' model (PFT only)			'Best' predictive model (PFTs, leaf traits and climate)				'Null' model (PFT only)		
Source	Value	s.e.	t-value	Value	s.e.	t-value	Source	Value	s.e.	t-value	Value	s.e.	t-value
PFT-BIT	1.2636	0.033	38.551	1.3805	0.046	29.750	PFT-BIT	8.5341	2.091	4.081	10.8938	1.243	8.764
PFT-C3H	0.4708	0.141	3.348	0.5099	0.160	3.185	PFT-C3H	-5.6273	6.832	-0.824	10.0926	3.569	2.828
PFT-NIT	-0.3595	0.150	-2.392	-0.0558	0.179	-0.311	PFT-NIT	6.8086	16.683	0.408	-2.2741	3.553	-0.640
PFT-S	0.3290	0.064	5.163	0.3460	0.071	4.867	PFT-S	-2.9249	2.564	-1.141	1.8429	1.492	1.235
[N] <sub>a</sub>	0.0728	0.018	4.124				[N] <sub>m</sub>	-0.1306	0.085	-1.531			
[P] <sub>a</sub>	0.0015	0.000	7.389				[P] <sub>m</sub>	-0.5670	1.491	-0.380			
$V_{\text{cmax,a}}^{25}$	0.0095	0.001	15.241				$M_a$	-0.0137	0.004	-3.040			
TWQ	-0.0358	0.006	-5.658				$V_{\text{cmax,m}}^{25}$	0.0111	0.002	6.459			
Interaction: C3H x [N] <sub>a</sub>	0.3394	0.069	4.892				Interaction: C3H x [N] <sub>m</sub>	0.7252	0.295	2.459			
Interaction: NIT x [N] <sub>a</sub>	0.0762	0.146	0.523				Interaction: NIT x [N] <sub>m</sub>	-0.7283	1.796	-0.405			
Interaction: S x [N] <sub>a</sub>	0.0687	0.053	1.295				Interaction: S x [N] <sub>m</sub>	0.1605	0.146	1.102			
							Interaction: C3H x [P] <sub>m</sub>	-4.2308	2.659	-1.591			
							Interaction: NIT x [P] <sub>m</sub>	0.4131	1.694	0.244			
							Interaction: S x [P] <sub>m</sub>	2.3333	1.790	1.303			
							Interaction: [N] <sub>m</sub> x [P] <sub>m</sub>	0.1876	0.062	3.026			
Random effects							Random effects						
'Best' predictive model (PFTs, leaf traits and climate)				'Null' model			'Best' predictive model (PFTs, leaf traits and climate)				'Null' model		
Source	# levels group <sup>-1</sup>	Residual variance	% of total	Residual variance	% of total		Source	# levels group <sup>-1</sup>	Residual variance	% of total	Residual variance	% of total	
Intercept variance: species	531	0.009	7.1%	0.023	11.5%		Intercept variance: families	100	0.373	0.7%	7.950	9.2%	
Intercept variance: families	100	0.002	1.4%	0.004	2.1%		Intercept variance: sites	49	37.745	73.2%	55.290	64.2%	
Intercept variance: sites	49	0.031	23.4%	0.073	36.2%		Residual error		13.476	26.1%	22.850	26.5%	
Residual error		0.091	68.2%	0.102	50.2%		Total		51.594	100.0%	86.090	100.0%	
Total		0.133	100.0%	0.202	100.0%								

For the 'best' models, parameter values, s.e. and t-values given for the continuous explanatory variables; explanatory variables (all un-transformed and centred on their means) are: (1) plant functional types (PFT), according to *JULES* (Clark *et al.*, 2011): BIT (broad-leaved tree), C3H ( $\text{C}_3$  metabolism herbs/grasses), NIT (needle-leaved trees), and S (shrubs); (2) leaf nitrogen ([N]) and phosphorus ([P]) concentrations ( $\text{g m}^{-2}$  for area-based values and  $\text{mg g}^{-1}$  for mass based values),  $M_a$  ( $\text{g m}^{-2}$ ) and Rubisco  $\text{CO}_2$  fixation capacity at  $25^\circ\text{C}$  ( $V_{\text{cmax}}^{25}$ ;  $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$  and  $\text{nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$  for area and mass-based values, respectively); and mean temperature of the warmest quarter (TWQ,  $^\circ\text{C}$ ) (Hijmans *et al.*, 2005). Figure S5 (Supp. Info.) assess heterogeneity and normality assumptions of the 'best' models, while Figure S6 shows model validation graphs for the area-based model fixed component explanatory variables; similarly, Fig. S7 shows details for variables omitted from the fixed components in the area-based model ( $M_a$ , AI and PWQ). The PFT-BIT values (first row) are based on the assumption that other variables were at their global mean values. The random effects tables, the intercept was allowed to vary among species, families and sites; residual errors shown are within species, families and sites. See Figure 7 for scatter plots of

modelled vs actual values of the 'best' models, both with and without inclusion of random effects. See also Table S4 (Supporting Information) for area-based model outputs for scenarios where different combinations of fixed effect parameters were included.

**Table 6. PFT-specific equations (formulated from the ‘best’ mixed-effects models shown in Table 5) that can be used to predict variability in (a) area-based ( $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ) and (b) mass-based ( $\text{nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$ ) leaf respiration at 25°C ( $R_{\text{dark,a}}^{25}$  and  $R_{\text{dark,m}}^{25}$ , respectively).**

(a) PFT-specific equations to predict variability in $R_{\text{dark,a}}^{25}$ (‘best’ model)	(b) PFT-specific equations to predict variability in $R_{\text{dark,m}}^{25}$ (‘best’ model)
BIT: $R_{\text{dark,a}}^{25} = 1.2636 + (0.0728*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax,a}}^{25}) - (0.0358*\text{TWQ})$	BIT: $R_{\text{dark,m}}^{25} = 8.5341 - (0.1306*[N]_m) - (0.5670*[P]_m) - (0.0137*M_a) + (0.0111*V_{\text{cmax,m}}^{25}) + (0.1876*([N]_m*[P]_m))$
C3H: $R_{\text{dark,a}}^{25} = 1.7344 + (0.4122*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax,a}}^{25}) - (0.0358*\text{TWQ})$	C3H: $R_{\text{dark,m}}^{25} = 2.9068 + (0.5946*[N]_m) - (4.7978*[P]_m) - (0.0137*M_a) + (0.0111*V_{\text{cmax,m}}^{25}) + (0.1876*([N]_m*[P]_m))$
NIT: $R_{\text{dark,a}}^{25} = 0.9041 + (0.1489*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax,a}}^{25}) - (0.0358*\text{TWQ})$	NIT: $R_{\text{dark,m}}^{25} = 15.3427 - (0.8589*[N]_m) - (0.1539*[P]_m) - (0.0137*M_a) + (0.0111*V_{\text{cmax,m}}^{25}) + (0.1876*([N]_m*[P]_m))$
S: $R_{\text{dark,a}}^{25} = 1.5926 + (0.1415*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax,a}}^{25}) - (0.0358*\text{TWQ})$	S: $R_{\text{dark,m}}^{25} = 5.6092 + (0.0299*[N]_m) + (1.7663*[P]_m) - (0.0137*M_a) + (0.0111*V_{\text{cmax,m}}^{25}) + (0.1876*([N]_m*[P]_m))$

Explanatory variables are: (1) plant functional types (PFT), according to *JULES* (Clark *et al.*, 2011): BIT (broad-leaved tree), C3H ( $\text{C}_3$  metabolism herbs/grasses), NIT (needle-leaved trees), and S (shrubs); (2) leaf nitrogen ([N]) and phosphorus ([P]) concentrations ( $\text{g m}^{-2}$  for area-based values and  $\text{mg g}^{-1}$  for mass based values), and Rubisco  $\text{CO}_2$  fixation capacity at 25°C ( $V_{\text{cmax}}^{25}$ ;  $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$  and  $\text{nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$  for area and mass-based values, respectively); and mean temperature of the warmest quarter (TWQ, °C) (Hijmans *et al.*, 2005). Note – equations refer to un-transformed values of each response and explanatory variable. See also Table S4 (Supporting Information) for area-based model equations for scenarios where different combinations of fixed effect parameters were included.

## Figure legends

**Figure 1. Location (a) and climate envelope (b) of the sites at which leaf dark respiration ( $R_{\text{dark}}$ ) and associated traits were measured.** (a) shows site locations on a global map showing spatial variability in mean annual temperatures (MAT); (b) shows plot of mean annual precipitation (MAP) vs MAT for each site (shown in biome classes). See Table 1 for summary of site information, and Table S1 and S2 (Supporting Information) for details on the latitude, longitude, altitude (height above sea level), MAT, mean temperature of the warmest quarter (i.e. warmest 3-month period per year; TWQ), MAP, mean precipitation of the warmest quarter (PWQ) and aridity index (AI, ratio of MAP to mean annual potential evapotranspiration). In (b), biomes categorization of each site is shown. Biome abbreviations: Tu, tundra; BF, boreal forest; TeDF, temperate deciduous forest; TeRF, temperate rainforest; TeW, temperate woodland; Sa, savana; TrRF\_up, upland tropical rainforest (>1500 m asl); TrRF\_low, lowland tropical rainforest (<1500 m asl). In (b), note the unusually high MAP at the Frans Josef TeRF site on the Sth Island of New Zealand.

**Figure 2. Box plots showing modulation of leaf structural and chemical traits by *JULES* (Clark *et al.*, 2011) plant functional type (PFT) classifications.** Traits shown are: (a)  $M_a$ , leaf mass per unit leaf area; (b)  $[N]_a$ , area-based leaf nitrogen concentration; and (c)  $[P]_a$ , area-based leaf phosphorous concentration. Data shown are for individual row observations contained in the *GlobResp* database (to give an indication of underlying data distribution). The central box in each box plot shows the interquartile range; the median is shown as the bold line within each box; whiskers extend 1.5 times the interquartile range or to the most extreme value, whichever is the smaller; any points outside these values are shown as individual points. Data for the following *JULES* (Clark *et al.*, 2011) plant functional type (PFT) classifications: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; C4H, C<sub>4</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub

**Figure 3. Box plots showing modulation of carboxylation capacity of Rubisco ( $V_{\text{cmax}}$ ) and leaf respiration ( $R_{\text{dark}}$ ) in darkness by *JULES* (Clark *et al.*, 2011) plant functional type (PFT) classifications.** Data shown are for individual row observations contained in the *GlobResp* database (to give an indication of underlying data distribution). Rates at 25°C are shown. Traits shown are: (a)  $V_{\text{cmax},a}^{25}$  and (c)  $V_{\text{cmax},m}^{25}$ : area- and mass-based carboxylation rates, respectively; (b)  $R_{\text{dark},a}^{25}$  and (d)  $R_{\text{dark},m}^{25}$ : area- and mass-based respiration rates, respectively. Values of  $V_{\text{cmax}}$  at 25°C were calculated according to Farquhar *et al.* (1980) assuming an activation energy ( $E_a$ ) of 64.8 kJ mol<sup>-1</sup>. Values of  $R_{\text{dark}}$  at 25°C were calculated assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). The central box in each box plot shows the interquartile range; the median is shown as the bold line within each box; whiskers extend 1.5 times the interquartile range or to the most extreme value, whichever is the smaller; any points outside these values are shown as individual points. Data for the following *JULES* (Clark *et al.*, 2011) plant functional type (PFT) classifications: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. Data not shown for C<sub>4</sub> metabolism herbs/grasses, due to limited data availability.



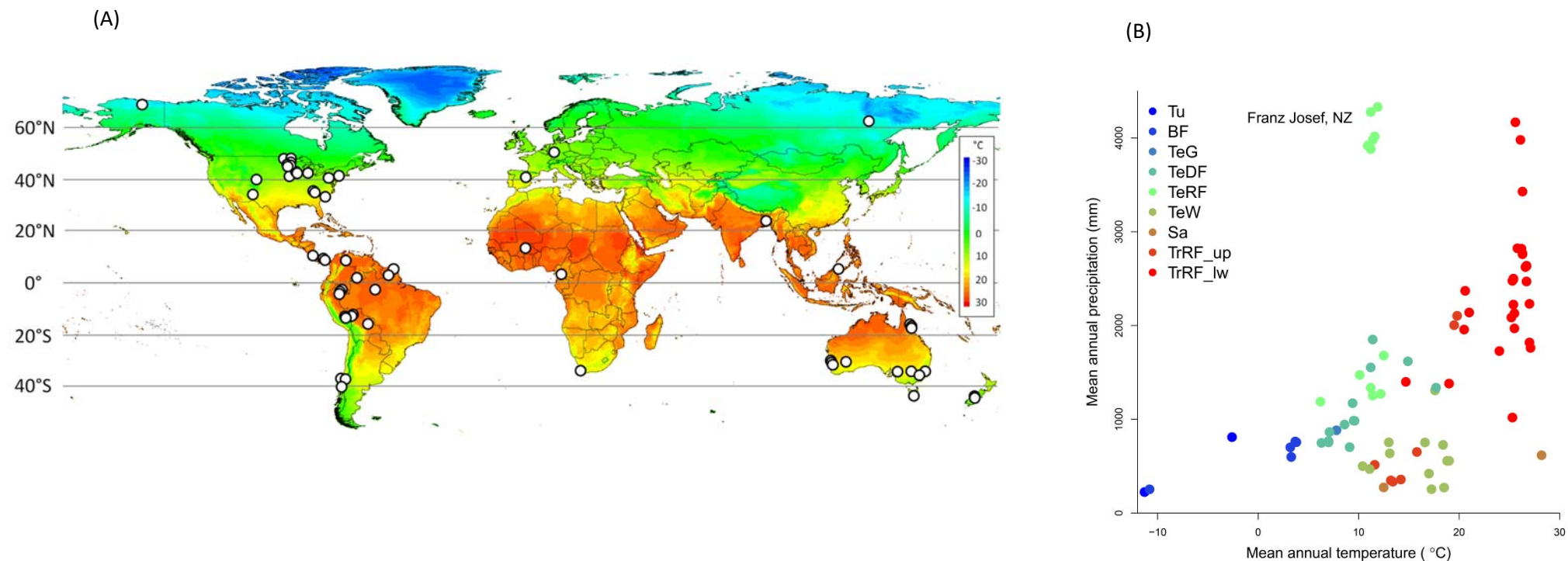
**Figure 4. Relationships between leaf  $R_{\text{dark}}$  (log<sub>10</sub> scale) and location (absolute latitude) or climate [mean daily temperature of the warmest quarter (TWQ) & aridity index (AI)].** Traits shown are:  $R_{\text{dark},a}^{25}$ , (a, b and c) and  $R_{\text{dark},a}^{\text{TWQ}}$  (d, e and f), predicted area-based  $R_{\text{dark}}$  rates at 25°C and TWQ, respectively;  $R_{\text{dark},m}^{25}$  (g, h and i) and  $R_{\text{dark},m}^{\text{TWQ}}$  (j, k and l), predicted mass-based  $R_{\text{dark}}$  rates at 25°C and TWQ, respectively. Values shown are averages for unique site:species combinations for rates at 25°C and TWQ, calculated assuming a temperature-dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). Values at the TWQ of each replicate were calculated using climate/location data from the *WorldClim* data base (Hijmans *et al.*, 2005). Aridity index calculated as the ratio of mean annual precipitation (MAP) to mean annual potential evapotranspiration (PET) (UNEP, 1997). In plots against latitude, northern and southern latitudes shown as blue and red symbols, respectively. Solid lines in each plot show regression lines where the relationships were significant; dashed lines show the prediction intervals (two-times the standard deviation) around the predicted relationship. See Table 3 for correlations between log<sub>10</sub> transformed  $R_{\text{dark}}$  and location/climate. Note: see Figure S3 (Supporting Information) for relationships between  $R_{\text{dark}}$  and AI, excluding data from the exceptionally high rainfall sites at Frans Josef on the Sth Island of New Zealand.

**Figure 5. Patterning of area- and mass-based  $R_{\text{dark}}^{25} - V_{\text{cmax}}^{25}$  relationships by *JULES* PFTs (a and d); TWQ categories (5°C intervals) – all data (b and e); and TWQ categories (5°C intervals) – broad-leaved trees only (c and f).** All values shown on a log<sub>10</sub> scale. Values shown are averages for unique site:species combinations. Upper panels (a, b and c) show area-based values, while lower panels (d, e and f) show mass-based values. *JULES* PFTs: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. TWQ classes: 1<sup>st</sup> <10°C; 2<sup>nd</sup> 10-15°C; 3<sup>rd</sup> 15-20°C; 4<sup>th</sup> 20-25°C; 5<sup>th</sup> >25°C. Values of  $R_{\text{dark}}$  at 25°C were calculated assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). Values  $V_{\text{cmax}}$  at 25°C were calculated according to Farquhar *et al.* (1980) assuming an activation energy ( $E_a$ ) of 64.8 kJ mol<sup>-1</sup>. See Table S3 for SMA regression outputs.

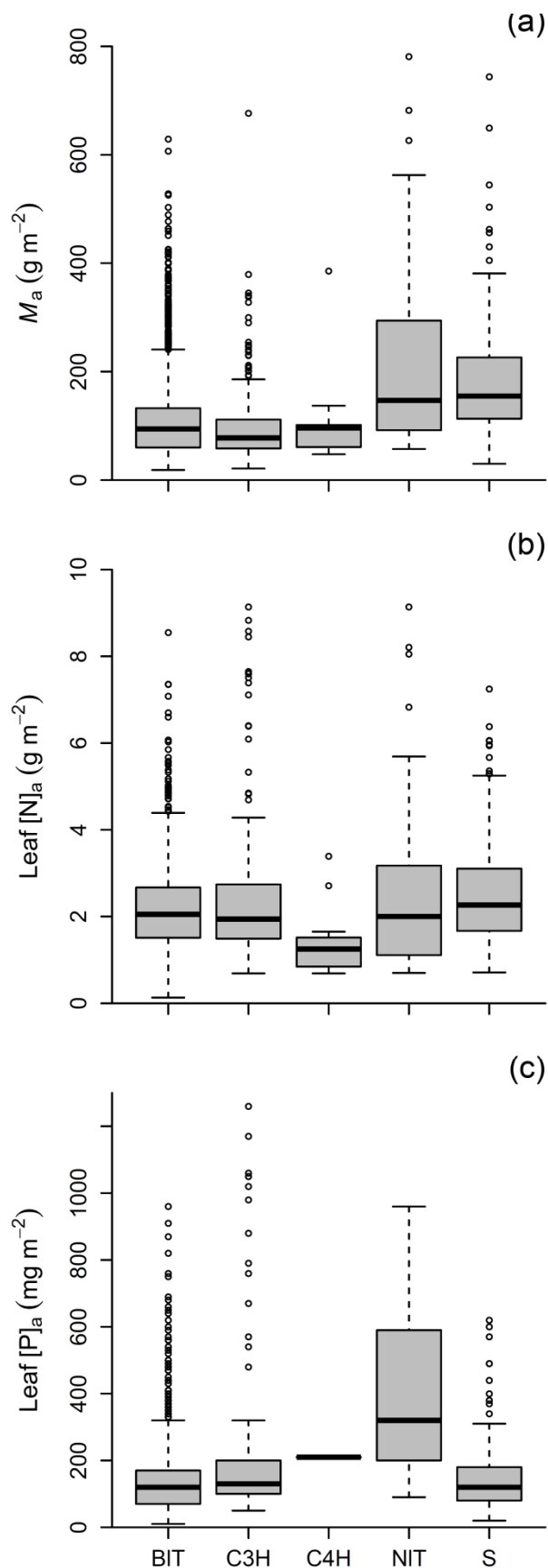
**Figure 6. Patterning of area- and mass-based  $R_{\text{dark}}^{25} - N$  relationships by *JULES* PFTs (a and d); TWQ categories (5°C intervals) – all data (b and e); and TWQ categories (5°C intervals) – broad-leaved trees only (c and f).** Values shown are averages for unique site:species combinations. All values shown on a log<sub>10</sub> scale. *JULES* PFTs: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. TWQ classes: 1<sup>st</sup> <10°C; 2<sup>nd</sup> 10-15°C; 3<sup>rd</sup> 15-20°C; 4<sup>th</sup> 20-25°C; 5<sup>th</sup> >25°C. Values of  $R_{\text{dark}}$  at 25°C were calculated assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). See Table S3 for SMA regression outputs.

**Figure 7. Scatterplots for (a) area-based, and (b) mass-based linear mixed-effects model's goodness of fits, including fixed and random terms.** Observed values of leaf respiration at 25°C ( $R_{\text{dark}}^{25}$ ) are plotted against model predictions (using the 'best' predictive models detailed in Table 5). For the area-based model (a), the fixed component explanatory variables were: (1) plant functional types (PFT), according to *JULES* (Clark *et al.*, 2011); (2) area-based leaf nitrogen ( $[N]_a$ ) and phosphorus ( $[P]_a$ ) concentrations, and Rubisco CO<sub>2</sub> fixation capacity at 25°C ( $V_{\text{cmax},a}^{25}$ ); and mean temperature of the warmest quarter (TWQ) (Hijmans *et al.*, 2005). For the mass-based model (b), the fixed component explanatory variables were: (1) plant functional types (PFT); (2) mass-based leaf nitrogen ( $[N]_m$ ) and phosphorus ( $[P]_m$ ) concentrations, Rubisco CO<sub>2</sub> fixation capacity at 25°C ( $V_{\text{cmax},m}^{25}$ ), and leaf mass per unit leaf area ( $M_a$ ).

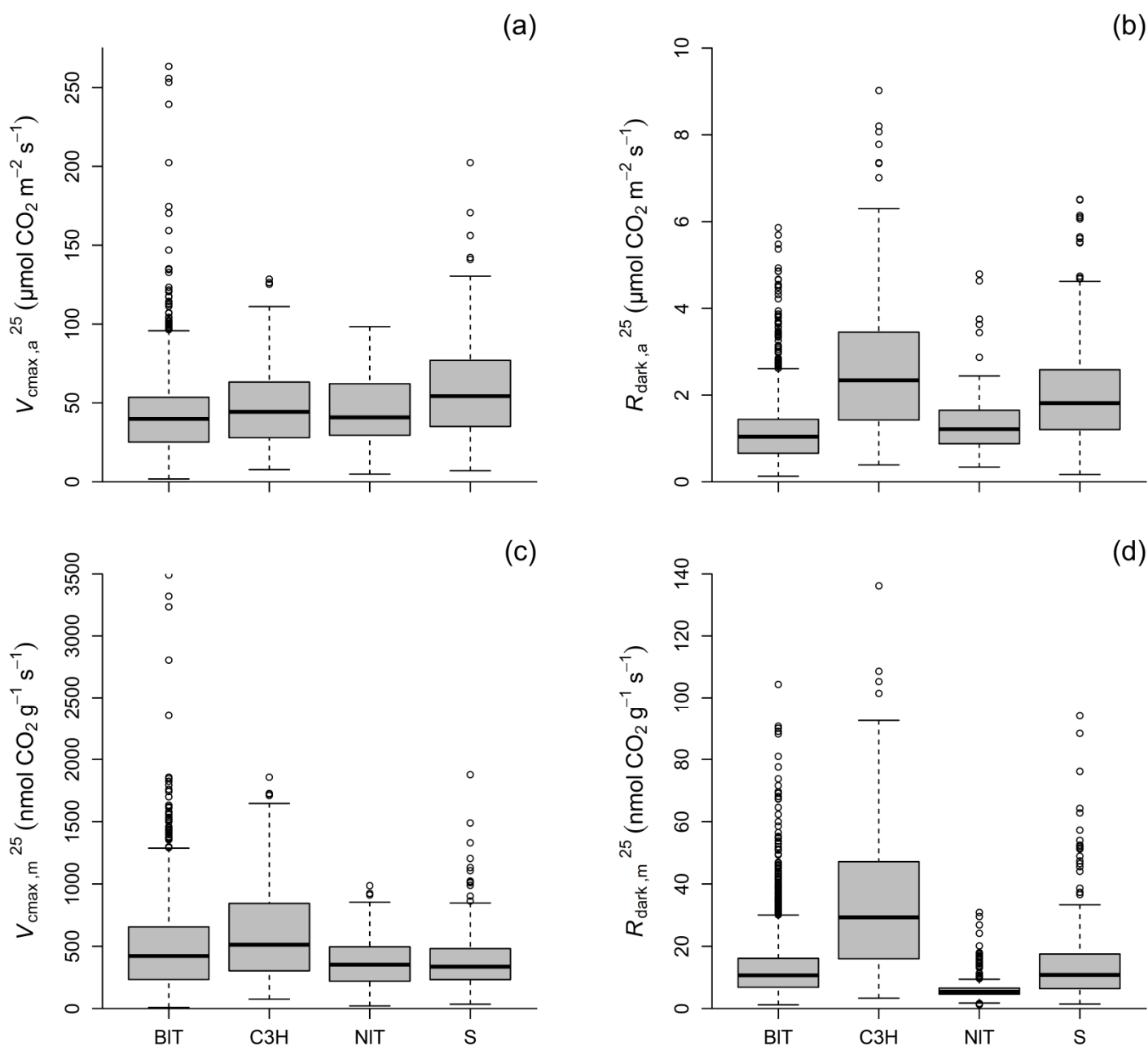
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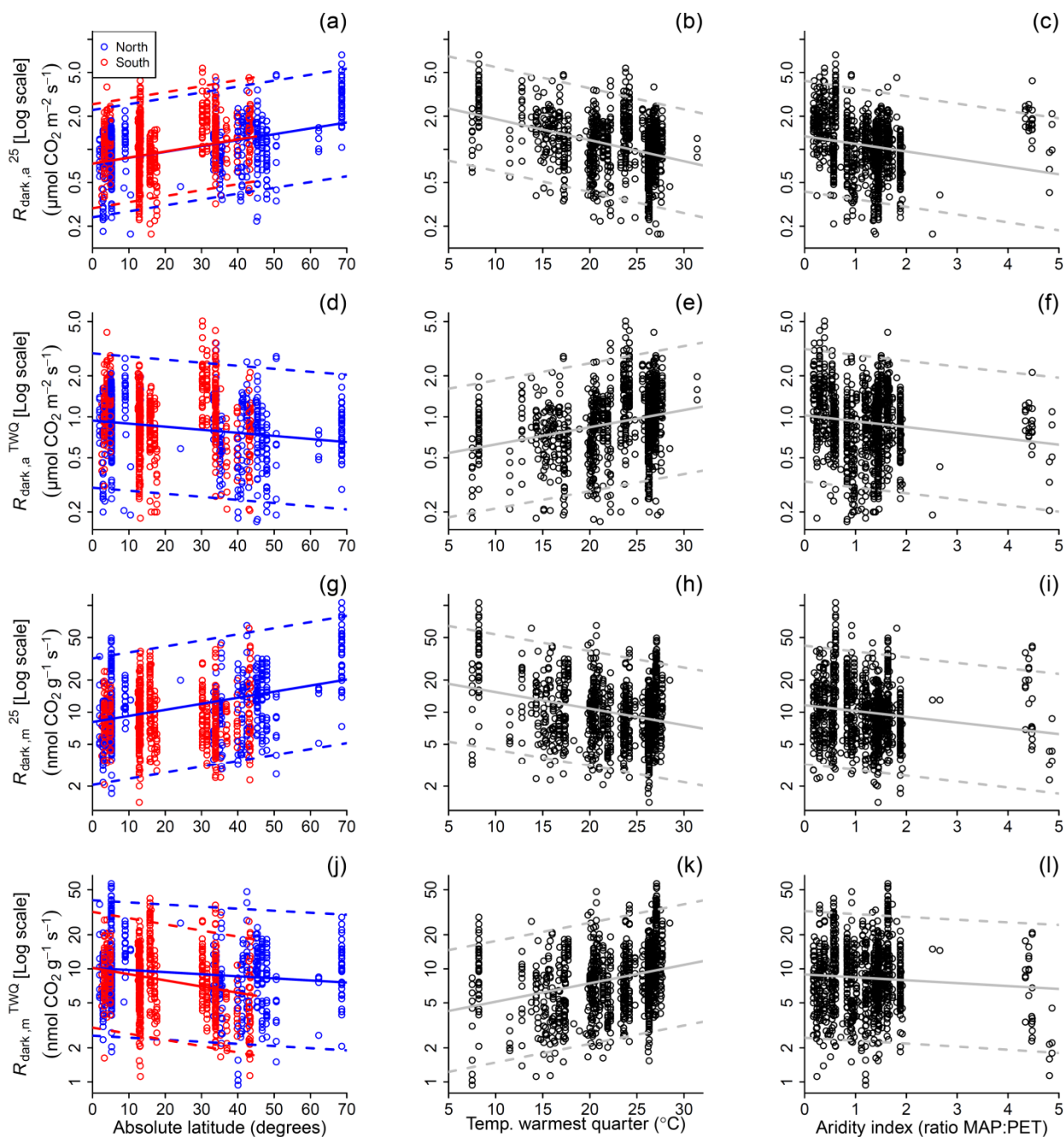
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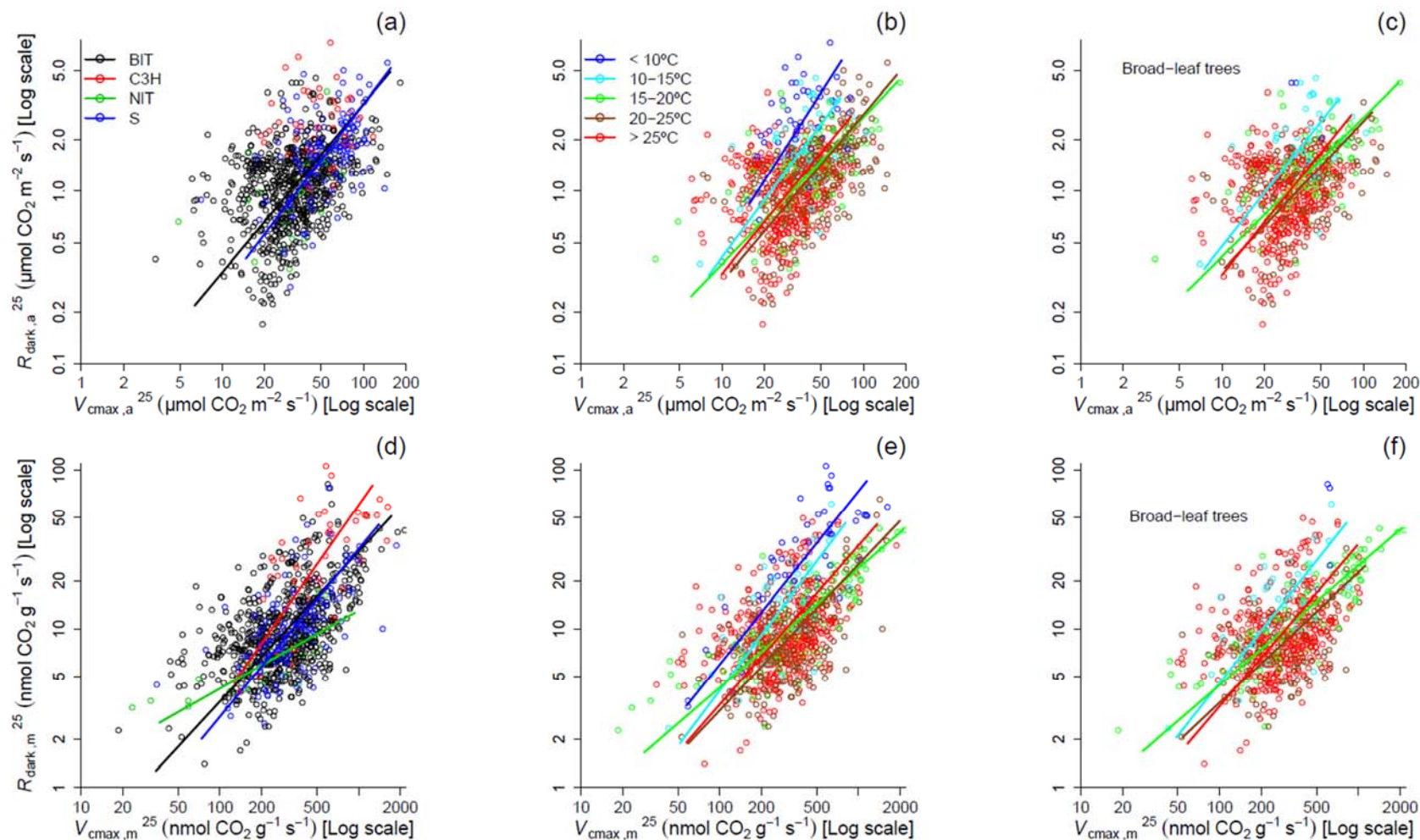
**Figure 3. Box plots showing modulation of carboxylation capacity of Rubisco ( $V_{\text{cmax}}$ ) and leaf respiration ( $R_{\text{dark}}$ ) in darkness by *JULES* (Clark *et al.*, 2011) plant functional type (PFT) classifications.** Data shown are for individual row observations contained in the *GlobResp* database (to give an indication of underlying data distribution). Rates at 25°C are shown. Traits shown are: (a)  $V_{\text{cmax},a}^{25}$  and (c)  $V_{\text{cmax},m}^{25}$ : area- and mass-based carboxylation rates, respectively; (b)  $R_{\text{dark},a}^{25}$  and (d)  $R_{\text{dark},m}^{25}$ : area- and mass-based respiration rates, respectively. Values of  $V_{\text{cmax}}$  at 25°C were calculated according to Farquhar *et al.* (1980) assuming an activation energy ( $E_a$ ) of 64.8 kJ mol<sup>-1</sup>. Values of  $R_{\text{dark}}$  at 25°C were calculated assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). The central box in each box plot shows the interquartile range; the median is shown as the bold line within each box; whiskers extend 1.5 times the interquartile range or to the most extreme value, whichever is the smaller; any points outside these values are shown as individual points. Data for the following *JULES* (Clark *et al.*, 2011) plant functional type (PFT) classifications: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. Data not shown for C<sub>4</sub> metabolism herbs/grasses, due to limited data availability.



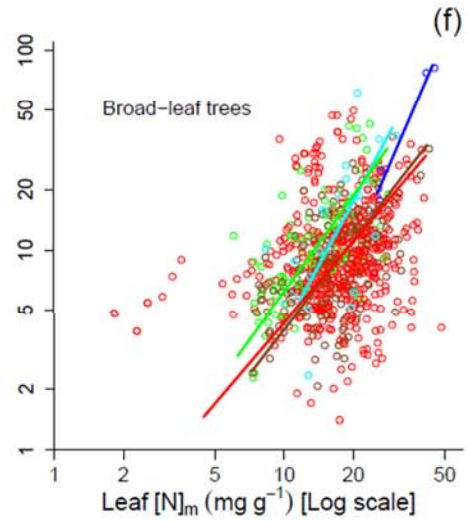
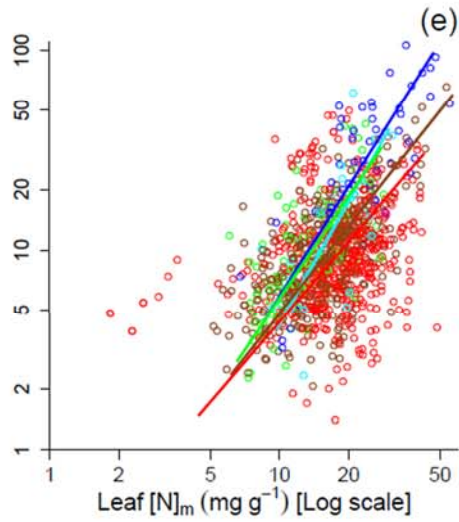
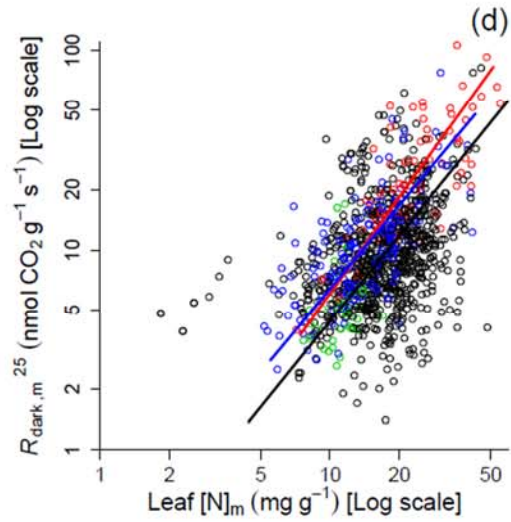
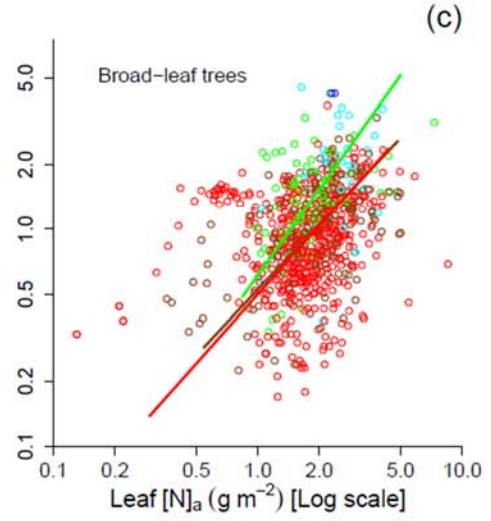
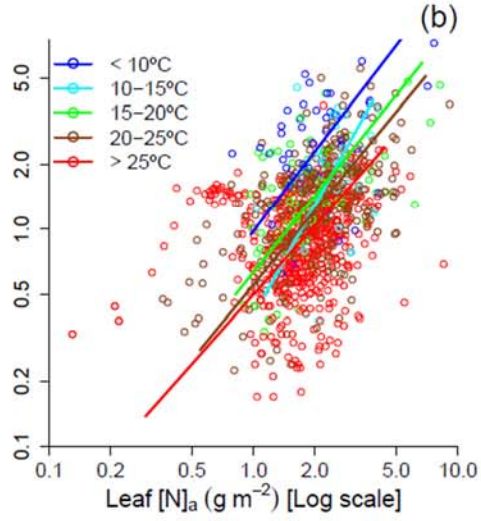
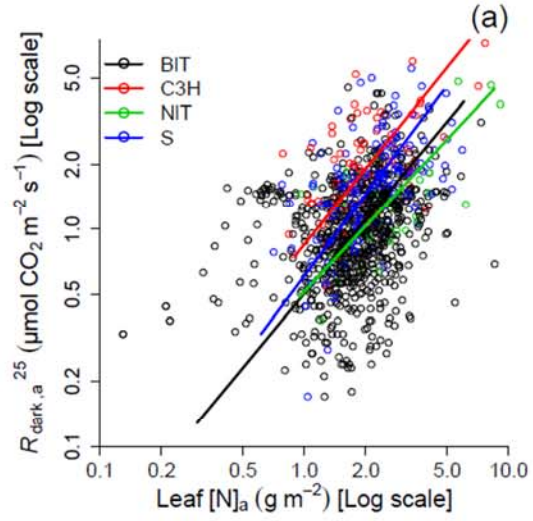
**Figure 4. Relationships between leaf  $R_{\text{dark}}$  ( $\log_{10}$  scale) and location (absolute latitude) or climate [mean daily temperature of the warmest quarter (TWQ) & aridity index (AI)].** Traits shown are:  $R_{\text{dark},a}^{25}$ , (a, b and c) and  $R_{\text{dark},a}^{\text{TWQ}}$  (d, e and f), predicted area-based  $R_{\text{dark}}$  rates at 25°C and TWQ, respectively;  $R_{\text{dark},m}^{25}$  (g, h and i) and  $R_{\text{dark},m}^{\text{TWQ}}$  (j, k and l), predicted mass-based  $R_{\text{dark}}$  rates at 25°C and TWQ, respectively. Values shown are averages for unique site:species combinations for rates at 25°C and TWQ, calculated assuming a temperature-dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). Values at the TWQ of each replicate were calculated using climate/location data from the *WorldClim* data base (Hijmans *et al.*, 2005). Aridity index calculated as the ratio of mean annual precipitation (MAP) to mean annual potential evapotranspiration (PET) (UNEP, 1997). In plots against latitude, northern and southern latitudes shown as blue and red symbols, respectively. Solid lines in each plot show regression lines where the relationships were significant; dashed lines show the prediction intervals (two-times the standard deviation) around the predicted relationship. See Table 3 for correlations between  $\log_{10}$  transformed  $R_{\text{dark}}$  and location/climate. Note: see Figure S3 (Supporting Information) for relationships between  $R_{\text{dark}}$  and AI, excluding data from the exceptionally high rainfall sites at Frans Josef on the Sth Island of New Zealand.



**Figure 5. Patterning of area- and mass-based  $R_{\text{dark}}^{25} - V_{\text{cmax}}^{25}$  relationships by *JULES* PFTs (a and d); TWQ categories (5°C intervals) – all data (b and e); and TWQ categories (5°C intervals) – broad-leaved trees only (c and f).** All values shown on a  $\log_{10}$  scale. Values shown are averages for unique site:species combinations. Upper panels (a, b and c) show area-based values, while lower panels (d, e and f) show mass-based values. *JULES* PFTs: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. TWQ classes: 1<sup>st</sup> <10°C; 2<sup>nd</sup> 10-15°C; 3<sup>rd</sup> 15-20°C; 4<sup>th</sup> 20-25°C; 5<sup>th</sup> >25°C. Values of  $R_{\text{dark}}$  at 25°C were calculated assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). Values  $V_{\text{cmax}}$  at 25°C were calculated according to Farquhar *et al.* (1980) assuming an activation energy ( $E_a$ ) of 64.8 kJ mol<sup>-1</sup>. See Table S3 for SMA regression outputs.

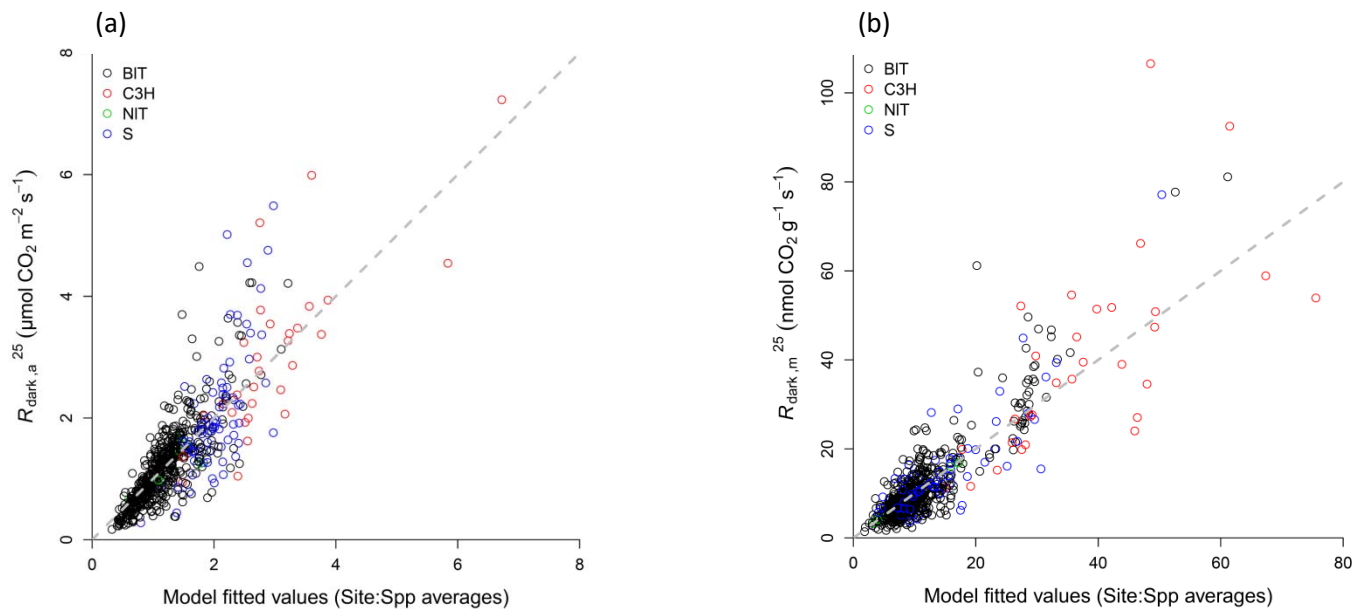


**Figure 6. Patterning of area- and mass-based  $R_{\text{dark}}^{25}$  – N relationships by *JULES* PFTs (a and d); TWQ categories (5°C intervals) – all data (b and e); and TWQ categories (5°C intervals) – broad-leaved trees only (c and f).** Values shown are averages for unique site:species combinations. All values shown on a  $\log_{10}$  scale. *JULES* PFTs: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. TWQ classes: 1<sup>st</sup> <10°C; 2<sup>nd</sup> 10-15°C; 3<sup>rd</sup> 15-20°C; 4<sup>th</sup> 20-25°C; 5<sup>th</sup> >25°C. Values of  $R_{\text{dark}}$  at 25°C were calculated assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). See Table S3 for SMA regression outputs.





**Figure 7. Scatterplots for (a) area-based, and (b) mass-based linear mixed-effects model's goodness of fits, including fixed and random terms.** Observed values of leaf respiration at 25°C ( $R_{\text{dark}}^{25}$ ) are plotted against model predictions (using the 'best' predictive models detailed in Table 5). For the area-based model (a), the fixed component explanatory variables were: (1) plant functional types (PFT), according to *JULES* (Clark *et al.*, 2011); (2) area-based leaf nitrogen ( $[N]_a$ ) and phosphorus ( $[P]_a$ ) concentrations, and Rubisco CO<sub>2</sub> fixation capacity at 25°C ( $V_{\text{cmax},a}^{25}$ ); and mean temperature of the warmest quarter (TWQ) (Hijmans *et al.*, 2005). For the mass-based model (b), the fixed component explanatory variables were: (1) plant functional types (PFT); (2) mass-based leaf nitrogen ( $[N]_m$ ) and phosphorus ( $[P]_m$ ) concentrations, Rubisco CO<sub>2</sub> fixation capacity at 25°C ( $V_{\text{cmax},m}^{25}$ ), and leaf mass per unit leaf area ( $M_a$ ).



## Supporting Information

**Authors: Atkin, Bloomfield, Reich, Tjoelker et al.**

**Title: Global variability in leaf respiration in relation to climate, plant functional types and leaf traits**

**Methods S1: Sampling methods and measurements protocols - unpublished data collected at sites detailed in Table S1.**

### *Species identification*

For work undertaken at the RAINFOR plots in Sth America (<http://www.rainfor.org/en/project/field-campaigns>), voucher specimens were collected and identified according to Lopez-Gonzalez *et al.* (2011). For Sth American plots associated with the Carnegie Institution *Spectranomics project* (<http://spectranomics.ciw.edu>), botanical vouchers were identified as detailed in Asner *et al.* (2014). Species identification at the TERN Supersites (<http://www.tern.org.au/Australian-SuperSite-Network-pg17873.html>) in Australia were identified by CSIRO, university and/or forest service botanical staff at each site.

### *Sampling method (1): Ex situ measurements made using cut branches*

Branches being sampled in the morning from the sun-facing upper canopy of individual plants; leaves had experienced at least two hours direct sunlight before branches were sampled. Branches were re-cut under water immediately after detachment. Thereafter, branches were transported to a nearby laboratory located for *ex situ* measurements of net CO<sub>2</sub> exchange.

### *Sampling method (2) In situ measurements using attached branches*

Leaf gas exchange measured using attached, sun-facing upper canopy leaves of individual plants, typically between 9 am and 1 pm for most sites, with the exception of measurements in Sth America, Siberia and Spain, where measurements were made upto 4 pm.

### Measurement methods – leaf gas exchange

(1) Measurements of respiration ( $R_{\text{dark}}$ ) and light-saturated photosynthesis under ambient [CO<sub>2</sub>] ( $A_{\text{sat}}$ ) and elevated [CO<sub>2</sub>] ( $A_{\text{max}}$ ): Most recent, fully expanded leaves were selected for measurement of net CO<sub>2</sub> exchange rates, using Licor 6400 Portable Photosynthesis Systems (Li-6400, LiCor, Lincoln, NE) using a 6 cm<sup>2</sup> leaf chamber with red-blue light source (6400-18 RGB Light Source, Licor, Lincoln, NE). Measurements were made at a relative humidity of 60-70%, and at the prevailing ambient day-time  $T$  of each site (6-41°C, depending on site location). Leaves were first exposed to saturating irradiance (1000 - 2000  $\mu\text{mol photons m}^{-2} \text{s}^{-1}$ , depending on species and site) and an reference line atmospheric [CO<sub>2</sub>] of 400 ppm for 10 minutes, after which rates of light-saturated net photosynthesis ( $A_{\text{sat}}$ ) was measured following equilibrium. Thereafter, atmospheric [CO<sub>2</sub>] was increased to 1500-2000 ppm (depending on site location), with CO<sub>2</sub>-saturated, light-saturated rates of net photosynthesis ( $A_{\text{max}}$ ) then being measured. Finally, which leaves were placed in darkness for 30-45 mins [to avoid post-illumination transients; (Azcón-Bieto & Osmond, 1983; Atkin *et al.*, 1998)] and rates of leaf respiration in darkness ( $R_{\text{dark}}$ ) measured. Flow rates

through the leaf chamber were set to 500 and 300  $\mu\text{mol s}^{-1}$  for measurements under light-saturation and darkness, respectively.

- (2) Measurements of  $R_{\text{dark}}$  and  $A_{\text{sat}}$ : As for (1), but without measurements made at saturating atmospheric  $[\text{CO}_2]$  (i.e. no estimate of  $A_{\text{max}}$ ).
- (3) Measurements of  $R_{\text{dark}}$  and  $A_{\text{sat}}$  (from A-I curves): As for (1), but with measurements of  $A_{\text{sat}}$  being limited to measurements made at an atmospheric  $[\text{CO}_2]$  of 400 ppm (i.e. no estimate of  $A_{\text{max}}$ ) as part of studies of the Kok-effect (Kok, 1948) using light-response curves of net  $\text{CO}_2$  exchange (Atkin *et al.*, 2013; Heskell *et al.*, 2014). Measurements commenced at 1800  $\mu\text{mol photons m}^{-2} \text{s}^{-1}$  and decreased to 1500, 100 and then at 5  $\mu\text{mol photons m}^{-2} \text{s}^{-1}$  intervals to darkness, where  $R_{\text{dark}}$  was measured. Measurements took place at the prevailing day-time air  $T$  at each site (RH 60-70%). An equilibrium period of two minutes was allowed at each irradiance level before net  $\text{CO}_2$  exchange was measured. During measurements,  $\text{CO}_2$  flow rates in the leaf cuvette were set to 500  $\mu\text{mol s}^{-1}$  for the measurements made at 1800  $\mu\text{mol photons m}^{-2} \text{s}^{-1}$  and 300  $\mu\text{mol s}^{-1}$  for those in darkness.

### Leaf area, mass and nutrient concentration measurements

At most sites, leaf area was typically determined on a 600 dots/inch flatbed top-illumination optical scanner, with area being quantified subsequently using *Image J* software (<http://imagej.nih.gov/ij/>). The scanned leaves were then dried at 70°C for a minimum of 72 h before dry mass (DM) was measured. Leaf mass per area was then calculated as grams DM  $\text{m}^2$ . For sites where both leaf N and P values were reported, concentrations of the two elements were determined with a LaChat QuikChem 8500 Series 2 Flow Injection Analysis System (Lachat Instruments, Milwaukee, WI, USA) using Kjeldahl acid digests (Allen, 1974). For sites where only leaf N was reported, samples were ground using a hammer mill (31–700 Hammer Mill; Glen Creston, Stanmore, UK), weighed into tin cups and combusted using a Carlo-Erba elemental analyser NA1500 (Thermo Fisher Scientific, Milan, Italy).

## Methods S2: Temperature normalization of respiration rates

To enable comparisons of leaf  $R_{\text{dark}}$ , we calculated rates both for a common temperature (i.e. 25°C) and the estimated growth  $T$  at each site (TWQ and MMT). To estimate rates of  $R_{\text{dark}}$  ( $R_2$ ) at a given  $T$  ( $T_2$ ), we calculated rates  $R_{\text{dark}}$  at 25°C ( $R_{\text{dark}}^{25}$ ), TWQ ( $R_{\text{dark}}^{\text{TWQ}}$ ) and MMT ( $R_{\text{dark}}^{\text{MMT}}$ ) assuming a fixed  $Q_{10}$  of 2.23 (Atkin *et al.*, 2005) using the equation:

$$R_2 = R_1 Q_{10}^{\left[\frac{T_2 - T_1}{10}\right]} \quad \text{Eqn 1}$$

where  $R_1$  represents the rate of  $R_{\text{dark}}$  at the measurement  $T$  ( $T_1$ ). This approach assumes that the  $Q_{10}$  remains constant across a range of leaf  $T$  - global surveys of the  $T$ -dependence of  $R_{\text{dark}}$  have shown, however, that the  $Q_{10}$  declines with increasing leaf  $T$  (Tjoelker *et al.*, 2001; Atkin & Tjoelker, 2003). Given this, we also calculated  $R_{\text{dark}}^{25}$ ,  $R_{\text{dark}}^{\text{TWQ}}$  and  $R_{\text{dark}}^{\text{MMT}}$  using a  $T$ -dependent  $Q_{10}$  (herein called ‘*var Q*<sub>10</sub>’) according to:

$$R_2 = R_1 \left( 3.09 - 0.043 \left[ \frac{(T_2 + T_1)}{2} \right] \right)^{\left[ \frac{T_2 - T_1}{10} \right]} \quad \text{Eqn 2}$$

Comparison of area-based rates of  $R_{\text{dark}}^{25}$  calculated using Eqns 1 and 2 revealed little overall difference in predicted rates at 25°C ( $r^2 = 0.995$ , Fig. S1). Estimates of  $R_{\text{dark}}^{\text{TWQ}}$  were likewise similar, irrespective of the equation used ( $r^2 = 0.991$ , Fig. S1). For subsequent analyses, we used Eqn 2 (i.e. *var Q*<sub>10</sub>) when estimating rates of  $R_{\text{dark}}^{25}$ ,  $R_{\text{dark}}^{\text{TWQ}}$  and  $R_{\text{dark}}^{\text{MMT}}$ .

**Table S1. Details on unpublished databases used in *GlobResp* database of leaf respiration ( $R_{\text{dark}}$ ). Shown are individual sample sites, climate and measurement conditions of the sites at which  $R_{\text{dark}}$  was measured. Sites shown in order from decreasing latitude from north to south. Data on climate are from the *WorldClim* data base (Hijmans *et al.*, 2005). Number of species, plants measured and *JULES* plant functional types (PFTs) at each site shown, according to: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; C4H, C<sub>4</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. Biome classes: BF, boreal forests; TeDF, temperate deciduous forest; TeG, temperate grassland; TeRF, temperate rainforest; TeW, temperate woodland; TrRF\_lw, lowland tropical rainforest (<1500 asl); TrRF\_up, upland tropical rainforest (>1500 asl); Tu, tundra. Abbreviations: mean temperature of the warmest quarter (i.e. warmest 3-month period per year; TWQ), mean annual precipitation (MAP), mean precipitation of the warmest quarter (PWQ), aridity index (AI) calculated as the ratio of MAP to mean annual potential evapotranspiration (UNEP, 1997; Zomer *et al.*, 2008). Australia-ACT, Australian Capital Territory; Australia-FNQ, Far North Queensland; Australia-TAS, Tasmania; Australia-WA, Western Australia; USA-AK, Alaska; USA-MN, Minnesota; USA-NY, New York; See Methods S1 text in Supporting Information for details on sampling methods and measurement protocols.**

Country/Region	Biome	Latitude	Longitude	Altitude (m asl)	MAT (°C)	TWQ (°C)	MAP (mm)	PWQ (mm)	AI	No. species	No. measurements	PFTs present	Sampling method (Methods S1)	Measurement method (Methods S2)	Primary person responsible for collection of unpublished data (& senior associate)
USA-AK	Tu	68.630	-149.600	720	-11.3	8.2	225	113	0.608	37	204	BIT, C3H, S	(1)	(3)	N. Mirotchnick (K. Griffin)
Russia-Siberia	BF	62.252	129.621	218	-10.8	15.4	254	122	0.458	3	40	BIT, NIT	(2)	(2)	J. Zaragoza-Castells (O. Atkin)
Russia-Siberia	BF	62.250	129.621	216	-10.8	15.4	254	122	0.458	2	30	BIT, NIT	(2)	(2)	J. Zaragoza-Castells (O. Atkin)
USA-MN	BF	47.944	-91.755	426	3.7	17.3	763	308	0.976	11	182	BIT, NIT	(1)	(2)	P. Reich
USA-MN	BF	46.704	-92.525	385	3.2	17.7	702	288	0.832	7	199	BIT	(1)	(2)	P. Reich
USA-MN	TeDF	45.169	-92.762	210	7.0	21.1	769	315	0.832	1	18	BIT	(1)	(2)	K. Sendall (P. Reich)
USA-NY	TeDF	41.420	-74.010	225	9.4	20.8	1,173	308	1.204	3	21	BIT	(1)	(3)	K. Griffin
USA-NY	TeDF	41.420	-74.010	225	9.4	20.8	1,173	308	1.204	3	18	BIT	(1)	(3)	K. Griffin
Spain	TeW	40.809	-2.237	980	10.4	18.9	501	102	0.496	1	28	BIT	(2)	(2)	J. Zaragoza-Castells (O. Atkin)
Spain	TeW	40.805	-2.227	1,060	11.1	19.6	471	95	0.464	1	24	BIT	(2)	(2)	J. Zaragoza-Castells (O. Atkin)
French Guiana	TrRF_lw	5.270	-52.920	21	25.8	26.2	2,824	222	1.881	43	65	BIT	(1)	(1)	J. Zaragoza-Castells (P. Meir)
French Guiana	TrRF_lw	5.270	-52.920	21	25.8	26.2	2,824	222	1.881	43	78	BIT	(1)	(1)	J. Zaragoza-Castells (P. Meir)
Peru-Amazon	TrRF_lw	-3.252	-72.908	111	20.6	21.4	2,371	676	1.401	20	20	BIT	(1)	(1)	Y. Ishida (J. Lloyd/O. Atkin)
Peru-Amazon	TrRF_lw	-3.256	-72.894	111	26.2	26.7	2,821	681	1.667	18	18	BIT	(1)	(1)	Y. Ishida (J. Lloyd/O. Atkin)
Peru-Amazon	TrRF_lw	-3.941	-73.440	120	26.3	26.8	2,769	711	1.637	14	14	BIT, S	(1)	(1)	Y. Ishida (J. Lloyd/O. Atkin)
Peru-Amazon	TrRF_lw	-3.949	-73.435	120	26.3	26.8	2,769	711	1.638	17	18	BIT	(1)	(1)	Y. Ishida (J. Lloyd/O. Atkin)
Peru-Amazon	TrRF_lw	-3.954	-73.427	120	26.3	26.8	2,762	708	1.633	22	22	BIT	(1)	(1)	Y. Ishida (J. Lloyd/O. Atkin)
Peru-Amazon	TrRF_lw	-4.878	-73.630	124	26.7	27.0	2,634	618	1.506	14	15	BIT	(1)	(1)	Y. Ishida (J. Lloyd/O. Atkin)
Peru-Amazon	TrRF_lw	-4.899	-73.628	124	26.7	27.0	2,639	620	1.506	18	18	BIT	(1)	(1)	Y. Ishida (J. Lloyd/O. Atkin)
Peru-Amazon	TrRF_lw	-12.534	-69.054	200	25.5	26.4	2,131	686	1.215	5	5	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Amazon	TrRF_lw	-12.830	-69.271	220	25.3	26.3	2,477	957	1.436	64	65	BIT	(1)	(1)	J. Zaragoza-Castells & R. Guerrieri
Peru-Amazon	TrRF_lw	-12.831	-69.284	220	25.4	26.3	2,491	961	1.445	8	8	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Amazon	TrRF_lw	-12.839	-69.296	200	25.4	26.3	2,501	964	1.452	71	75	BIT	(1)	(1)	J. Zaragoza-Castells & R. Guerrieri (P. Meir/O. Atkin)
Peru-Andes	TrRF_up	-13.047	-71.542	1,750	19.5	20.3	2,005	574	1.196	17	20	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Andes	TrRF_up	-13.049	-71.537	1,500	20.6	21.4	2,371	676	1.402	14	16	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Andes	TrRF_up	-13.070	-71.556	1,800	19.8	20.6	2,104	602	1.249	20	20	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)

Peru-Andes	TrRF_up	-13.106	-71.589	2,750	15.8	16.8	652	188	0.423	10	11	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Andes	TrRF_up	-13.109	-71.600	3,000	14.2	15.3	359	103	0.244	8	8	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Andes	TrRF_up	-13.114	-71.607	3,450	11.6	12.8	515	160	0.367	13	14	BIT, C3H	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Andes	TrRF_up	-13.176	-71.595	3,000	13.2	14.3	349	101	0.24	14	16	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Peru-Andes	TrRF_up	-13.191	-71.588	3,000	13.4	14.5	335	97	0.23	7	7	BIT	(1)	(1)	R. Guerrieri (P. Meir/O. Atkin)
Australia-FNQ	TrRF_lw	-17.109	145.603	818	20.5	23.3	1,958	886	1.35	6	15	BIT	(1)	(3)	J. Zaragoza-Castells (O. Atkin/P. Meir)
Australia-FNQ	TrRF_lw	-17.120	145.632	728	21.0	23.8	2,140	954	1.471	16	56	BIT	(1)	(1)	L. Weerasinghe (O. Atkin)
Australia-FNQ	TrRF_lw	-17.682	145.534	1,040	19.0	22.2	1,382	641	0.943	10	24	BIT, S	(1)	(3)	J. Zaragoza-Castells(O. Atkin/P. Meir)
Australia-WA	TeW	-30.180	115.000	90	19.0	23.9	558	33	0.386	8	31	BIT, C3H, S	(2)	(1)	L. Weerasinghe (O. Atkin)
Australia-WA	TeW	-30.240	115.070	23	18.8	23.8	558	35	0.389	10	39	BIT, S	(2)	(1)	L. Weerasinghe (O. Atkin)
Australia-WA	TeW	-30.240	115.060	5	18.8	23.8	558	35	0.389	9	34	BIT, C3H, S	(2)	(1)	L. Weerasinghe (O. Atkin)
Australia-WA	TeW	-30.264	120.692	459	18.5	25.6	273	64	0.177	9	87	BIT, S	(1), (2)	(1)	K. Bloomfield (O. Atkin)
Australia-SA	TeW	-34.037	140.674	35	17.3	23.6	255	52	0.172	10	78	BIT, C3H, S	(1), (2)	(1)	K. Bloomfield (O. Atkin)
Australia-ACT	TeW	-35.276	149.109	601	13.1	19.8	637	162	0.509	5	18	BIT, S	(1), (2)	(3)	K. Crous (O. Atkin)
Australia-TAS	TeRF	-43.089	146.651	217	10.1	13.8	1,474	237	1.813	3	13	BIT	(1)	(1)	L. Weerasinghe (O. Atkin)
Australia-TAS	TeRF	-43.092	146.684	257	11.2	14.8	1,338	212	1.648	2	6	BIT, S	(1)	(1)	L. Weerasinghe (O. Atkin)
Australia-TAS	TeRF	-43.095	146.724	88	11.4	15.1	1,255	199	1.463	9	29	BIT, S	(1)	(1)	L. Weerasinghe (O. Atkin)

**Table S2. Details on published databases used in *GlobResp* database of leaf respiration ( $R_{\text{dark}}$ ). Shown are climate and measurement conditions of the sites at which  $R_{\text{dark}}$  was measured.** Sites shown in order from decreasing latitude from north to south. Data on climate are from the *WorldClim* data base (Hijmans *et al.*, 2005). Number of species and *JULES* plant functional types (PFTs) at each site shown, according to: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; C4H, C<sub>4</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. Biome classes: BF, boreal forests; TeDF, temperate deciduous forest; TeG, temperate grassland; TeRF, temperate rainforest; TeW, temperate woodland; TrRF\_lw, lowland tropical rainforest (<1500 asl); Tu, tundra. Abbreviations: mean temperature of the warmest quarter (i.e. warmest 3-month period per year; TWQ), mean annual precipitation (MAP), mean precipitation of the warmest quarter (PWQ), aridity index (AI) calculated as the ratio of MAP to mean annual potential evapotranspiration (UNEP, 1997; Zomer *et al.*, 2008). Australia-ACT, Australian Capital Territory; Australia-FNQ, Far North Queensland; Australia-NSW, New South Wales; Australia-WA, Western Australia; USA-AK, Alaska; USA-CO, Colorado; USA-MN, Minnesota; USA-IW, Iowa; USA-WI, Wisconsin; USA-MI, Michigan; USA-PN, Pennsylvania; USA-NC, North Carolina; USA-KT, Kentucky; USA-TN, Tennessee; USA-NM, New Mexico; USA-SC, South Carolina.

Country/Region	Biome	Latitude	Longitude	Altitude (m asl)	MAT (°C)	TWQ (°C)	MAP (mm)	PWQ (mm)	AI	No. species	PFTs present	Traits available in <i>GlobResp</i> database	References/Source
Germany	TeDF	50.600	8.700	60	9.1	17.2	704	190	0.917	9	BIT, NIT	$R_{\text{dark}}$ , [N], $M_a$	Grueters (1998); Kattge <i>et al.</i> (2011)
USA-MN	BF	47.803	-95.007	400	3.3	18.3	599	278	0.749	1	NIT	$R_{\text{dark}}$ , [N]	Tjoelker <i>et al.</i> (2008)
USA-MN	BF	46.721	-92.457	380	3.8	17.4	757	304	0.906	7	BIT	$R_{\text{dark}}$ , [N]	Machado & Reich (2006)
USA-MN	BF	46.705	-92.525	380	3.7	17.4	764	308	0.905	7	BIT, NIT	$R_{\text{dark}}$ , [N]	Tjoelker <i>et al.</i> (2008); Reich <i>et al.</i> (2008)
USA-MN	TeG	45.410	-93.210	300	6.3	20.4	749	314	0.835	35	BIT, C3H, C4H, S	$A_{\text{sat}}$ , $C_i$ , $R_{\text{dark}}$ , [N], $M_a$	Craine <i>et al.</i> (1999); Tjoelker <i>et al.</i> (2005)
USA-MN	TeDF	45.410	-93.210	300	6.3	20.4	749	314	0.835	3	BIT	$A_{\text{sat}}$ , $C_i$ , $R_{\text{dark}}$ , $M_a$	Tjoelker <i>et al.</i> (2005); Sendall & Reich (2013)
USA-MN	TeDF	44.996	-93.189	281	7.0	21.0	755	314	0.835	3	BIT	$R_{\text{dark}}$ , [N], $M_a$	Lee <i>et al.</i> (2005); Kattge <i>et al.</i> (2011)
USA-WI	TeDF	42.980	-90.120	360	7.1	20.2	865	315	0.932	1	BIT	$A_{\text{sat}}$ , $R_{\text{dark}}$ , [N], $M_a$	Reich <i>et al.</i> (1998b)
USA-MI	TeDF	42.530	-85.855	200	8.6	19.9	944	268	0.98	1	NIT	$R_{\text{dark}}$ , [N]	Tjoelker <i>et al.</i> (2008); Reich <i>et al.</i> (2008)
USA-WI	TeG	42.500	-90.000	275	7.8	20.7	884	315	0.925	15	BIT, C3H, NIT	$A_{\text{sat}}$ , $C_i$ , $R_{\text{dark}}$ , [N], $M_a$	Reich <i>et al.</i> (1998a); Reich <i>et al.</i> (1998b)
USA-IA	TeDF	41.170	-92.870	385	7.1	20.2	865	315	0.834	11	BIT, NIT	$R_{\text{dark}}$ , [N], $M_a$	Lusk & Reich (2000)
USA-PA	TeDF	40.82	-77.93	400	9.1	17.2	704	190	0.71	1	BIT	$A_{\text{sat}}$ , $R_{\text{dark}}$ , $M_a$	Kloeppel <i>et al.</i> (1993; 1994)
USA-PA	TeDF	40.8	-77.83	335	9.6	20.8	984	286	0.972	2	BIT	$A_{\text{sat}}$ , $C_i$ , $R_{\text{dark}}$ , [N], $M_a$	Kloeppel & Abrams (1995)
USA-PA	TeDF	40.78	-77.88	348	9.5	20.6	986	285	0.986	1	BIT	$A_{\text{sat}}$ , $C_i$ , $R_{\text{dark}}$ , [N], $M_a$	Kloeppel & Abrams (1995)
USA-CO	Tu	40.050	-105.600	3,360	-2.6	7.5	811	203	1.198	10	BIT, C3H, NIT, S	$A_{\text{sat}}$ , $C_i$ , $R_{\text{dark}}$ , [N], $M_a$	Reich <i>et al.</i> (1998b)
Japan	TeDF	35.720	140.800	20	14.9	23.7	1,619	433	1.921	4	BIT	$A_{\text{sat}}$ , $R_{\text{dark}}$ , [N], $M_a$	Miyazawa <i>et al.</i> (1998)
USA-TN	TeDF	35.500	-83.500	775	11.2	20.1	1,554	389	1.335	13	BIT, C3H, NIT, S	$A_{\text{sat}}$ , $R_{\text{dark}}$ , [N], $M_a$	Bolstad <i>et al.</i> (1999)
USA-NC	TeDF	35.050	-83.420	850	11.4	20.0	1,852	444	1.521	15	BIT, NIT	$R_{\text{dark}}$ , [N], $M_a$	Mitchell <i>et al.</i> (1999); Reich <i>et al.</i> (1998b)
USA-NM	Sa	34.000	-107.000	1,620	12.5	22.2	275	127	0.189	9	BIT, NIT, S	$A_{\text{sat}}$ , $C_i$ , $R_{\text{dark}}$ , [N], $M_a$	Reich <i>et al.</i> (1998b)
USA-SC	TeDF	33.330	-79.220	3	17.7	25.8	1,339	469	1.02	10	BIT, C3H, NIT, S	$R_{\text{dark}}$ , [N], $M_a$	Reich <i>et al.</i> (1998a; 1999)

Bangladesh	TrRF_lw	24.200	90.150	21	25.5	28.5	1,970	736	1.344	1	BIT	$A_{\text{sat}}, R_{\text{dark}}, M_a$	Kamaluddin & Grace (1993)
Niger	Sa	13.200	-2.230	280	28.2	31.4	618	55	0.304	3	BIT, S	$A_{\text{sat}}, R_{\text{dark}}$	Meir <i>et al.</i> (2007)
Costa Rica	TrRF_lw	10.470	-84.030	140	25.6	26.6	4,168	750	2.658	1	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, M_a$	Oberbauer & Strain (1985); (1986)
Costa Rica	TrRF_lw	10.430	-83.980	105	26.1	27.2	3,981	731	2.515	1	S	$A_{\text{sat}}, R_{\text{dark}}, [N], M_a$	Chazdon & Kaufmann (1993)
Panama	TrRF_lw	9.170	-79.850	90	26.6	27.5	2,624	410	1.877	1	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Zotz & Winter (1996)
Panama	TrRF_lw	8.983	-79.550	100	27.0	27.7	1,820	300	1.186	13	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Slot <i>et al.</i> (2014b)
Panama	TrRF_lw	8.970	-79.530	30	27.1	27.7	1,762	290	1.143	6	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Kitajima <i>et al.</i> (1997)
Venezuela	TrRF_lw	8.650	-71.400	2,350	14.7	15.1	1,400	458	1.053	1	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	García-Núñez <i>et al.</i> (1995)
Malaysia	TrRF_lw	5.160	117.900	20	26.7	27.1	2,471	501	1.638	29	Malaysia-Borneo	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Swaine (2007)
Cameroon	TrRF_lw	3.380	11.500	550	24.0	24.8	1,729	417	1.126	6	Cameroon	$A_{\text{sat}}, R_{\text{dark}}, [N], M_a$	Meir <i>et al.</i> (2007)
Suriname	TrRF_lw	2.854	-54.958	215	25.4	26.3	2,224	165	1.365	25	Suriname	$A_{\text{sat}}, R_{\text{dark}}, [N], M_a$	Kattge <i>et al.</i> (2011)
Venezuela	TrRF_lw	1.930	-67.050	120	26.3	26.6	3,430	740	1.725	9	Venezuela	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Reich <i>et al.</i> (1998b)
Brazil-Amazon	TrRF_lw	-2.580	-60.100	115	27.0	27.6	2,232	401	1.385	9	BIT	$R_{\text{dark}}, [N], M_a$	Meir <i>et al.</i> (2002)
Bolivia	TrRF_lw	-15.783	-62.917	400	25.3	27.0	1,020	436	0.57	50	BIT	$A_{\text{sat}}, R_{\text{dark}}, [N], M_a$	Poorter & Bongers (2006)
Australia-FNQ	TrRF_lw	-16.100	145.450	90	25.2	27.5	2,087	1,031	1.393	18	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Weerasinghe <i>et al.</i> (2014)
Australia-WA	TeW	-31.500	115.690	15	18.4	23.6	728	39	0.541	25	BIT, C3H, S	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Wright <i>et al.</i> (2004)
Sth Africa	TeW	-33.830	18.830	600	16.6	21.0	754	67	0.572	5	BIT, S	$A_{\text{sat}}, R_{\text{dark}}, [N], M_a$	Mooney <i>et al.</i> (1983)
Australia-NSW	TeW	-33.840	145.880	223	17.0	24.2	422	98	0.294	19	BIT, C3H, NIT, S	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Wright <i>et al.</i> (2001)
Australia-NSW	TeW	-33.840	145.880	223	17.0	24.2	422	98	0.294	21	BIT, C4H, S	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Wright <i>et al.</i> (2001)
Australia-NSW	TeW	-33.860	151.210	39	17.6	21.9	1,309	358	NA	18	BIT, S	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Wright <i>et al.</i> (2001)
Australia-NSW	TeW	-33.860	151.210	39	17.6	21.9	1,309	358	NA	17	BIT, S	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Wright <i>et al.</i> (2001)
Australia-ACT	TeW	-35.312	149.058	560	13.0	21.0	755	314	0.601	1	NIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Reich <i>et al.</i> (1999))
Chile	TeRF	-36.840	-73.030	30	12.2	16.1	1,272	74	1.208	6	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Wright <i>et al.</i> (2006)
Chile	TeRF	-37.000	-71.470	1,000	6.2	11.5	1,189	74	1.119	5	BIT, NIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Wright <i>et al.</i> (2006)
Chile	TeRF	-39.800	-73.000	400	12.5	16.7	1,680	129	1.622	12	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], M_a$	Wright <i>et al.</i> (2006)
New Zealand	TeRF	-43.250	170.180	68	11.9	16.3	4,331	1,103	4.866	3	BIT, NIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Atkin <i>et al.</i> (2013)
New Zealand	TeRF	-43.310	170.170	143	11.2	15.8	4,277	1,076	4.816	3	BIT, NIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Atkin <i>et al.</i> (2013)
New Zealand	TeRF	-43.380	170.180	134	11.6	16.2	4,017	1,017	4.468	3	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Atkin <i>et al.</i> (2013)
New Zealand	TeRF	-43.400	170.170	234	11.4	16.0	3,980	1,004	4.477	7	BIT	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Atkin <i>et al.</i> (2013)



New Zealand	TeRF	-43.410	170.170	271	10.9	15.6	3,920	980	4.409	6	BIT, S	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Atkin <i>et al.</i> (2013)
New Zealand	TeRF	-43.420	170.170	215	11.2	15.8	3,883	976	4.343	5	BIT, S	$A_{\text{sat}}, C_i, R_{\text{dark}}, [N], [P], M_a$	Atkin <i>et al.</i> (2013)

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**Table S3. Standardized Major Axis regression slopes and their confidence intervals for log-log transformed relationships shown in Figures 5 and 6 in the main text.** Coefficients of determination ( $r^2$ ) and significance values ( $p$ ) of each bivariate relationship are shown. 95% confidence intervals (CI) of SMA slopes and y-axis intercepts are shown in parentheses. In cases where SMA tests for common slopes revealed no significant differences between the upper canopy and lower canopy groups (*i.e.*  $P > 0.05$ ), when plotting bivariate relationships, common slopes were used (with CI of the common slopes provided). Where there was a significant difference in elevation of the common-slope SMA regressions, values for the y-axis intercept (elevation) are provided. Where appropriate, significant shifts along a common slopes are indicated. JULES PFTs: BIT, broad-leaved tree; C3H, C<sub>3</sub> metabolism herb/grass; C4H, C<sub>4</sub> metabolism herb/grass; NIT, needle-leaved tree; S, shrub. TWQ classes: <10°C; 10-15°C; 15-20°C; 20-25°C; >25°C. *Abbreviations:*  $R_{\text{dark,a}}^{25}$ , predicted area-based  $R_{\text{dark}}$  at 25°C;  $R_{\text{dark,m}}^{25}$ , mass-based  $R_{\text{dark}}$  at 25°C;  $V_{\text{cmax,a}}^{25}$ , predicted area-based  $V_{\text{cmax}}$  at 25°C;  $V_{\text{cmax,m}}^{25}$ , predicted mass-based  $V_{\text{cmax}}$  at 25°C

Figure	Response	Bivariate	JULES PFTs	H0 #1: No difference in slope (p-value)	PFT or TWQ-class (°C)	n	r <sup>2</sup>	p	Slope	Pairwise comparison	Slope CI_low	Slope CI_high	Intercept	H0 #2: No difference in elevation (p-value)	Intercepts for a common slope	Pairwise comparison (where relationship p significant)	H0 #3: No difference in 'shift' p-value
5(a)	$R_{\text{dark,a}}^{25}$	$V_{\text{cmax,a}}^{25}$	All bar C4H	0.7017	BIT	691	0.12	< 0.0001	0.976		0.910	1.046	-1.445	< 0.0001	-1.470	a	< 0.0001
					C3H	45	0.00	<b>0.8940</b>	1.073		0.793	1.453	-1.414				
					NIT	23	0.16	<b>0.0578</b>	0.949		0.633	1.422	-1.445				
					S	115	0.16	< 0.0001	1.076		0.908	1.276	-1.647				
5(d)	$R_{\text{dark,m}}^{25}$	$V_{\text{cmax,m}}^{25}$	All bar C4H	< 0.0001	BIT	682	0.27	< 0.0001	0.946	b	0.887	1.009	-1.351	< 0.0001	-1.279	a	< 0.0001
					C3H	44	0.37	< 0.0001	1.247	a	0.977	1.592	-1.962				
					NIT	23	0.62	< 0.0001	0.494	c	0.375	0.651	-0.366				
					S	115	0.31	< 0.0001	1.057	a, b	0.906	1.234	-1.671				
5(b)	$R_{\text{dark,a}}^{25}$	$V_{\text{cmax,a}}^{25}$	All bar C4H	0.0857	< 10	47	0.19	0.0023	1.273		0.974	1.662	-1.592	< 0.0001	-1.134	d	< 0.0001
					10 to 15	43	0.18	0.0042	1.103		0.832	1.461	-1.484				
					15 to 20	121	0.33	< 0.0001	0.849		0.732	0.985	-1.270				
					20 to 25	263	0.30	< 0.0001	0.966		0.872	1.069	-1.487				
					> 25	400	0.03	0.0004	0.999		0.907	1.101	-1.475				
5(e)	$R_{\text{dark,m}}^{25}$	$V_{\text{cmax,m}}^{25}$	All bar C4H	< 0.0001	< 10	47	0.62	< 0.0001	1.093	a	0.909	1.314	-1.412	< 0.0001	-1.287	c	< 0.0001
					10 to 15	42	0.38	< 0.0001	1.165	a	0.908	1.496	-1.720				
					15 to 20	121	0.68	< 0.0001	0.752	b	0.679	0.832	-0.875				
					20 to 25	258	0.31	< 0.0001	0.920	a	0.831	1.019	-1.356				
					> 25	396	0.15	< 0.0001	1.002	a	0.914	1.098	-1.482				
5(c)	$R_{\text{dark,a}}^{25}$	$V_{\text{cmax,a}}^{25}$	BIT only	0.0480	< 10	4	0.63	<b>0.2070</b>	-2.446		-9.686	-0.618	4.306	< 0.0001	-1.061		< 0.0001
					10 to 15	39	0.21	0.0036	1.033		0.771	1.384	-1.352				
					15 to 20	101	0.35	< 0.0001	0.805		0.685	0.945	-1.183				
					20 to 25	152	0.17	< 0.0001	0.865		0.747	1.001	-1.325				
					> 25	395	0.03	0.0006	1.011		0.917	1.115	-1.494				
5(f)	$R_{\text{dark,m}}^{25}$	$V_{\text{cmax,a}}^{25}$	BIT only	< 0.0001	< 10	4	0.41	<b>0.3627</b>	8.035		1.642	39.317	-20.639	< 0.0001	-1.061		< 0.0001
					10 to 15	39	0.40	< 0.0001	1.103	a	0.855	1.423	-1.549				
					15 to 20	101	0.72	< 0.0001	0.753	b	0.678	0.836	-0.862				
					20 to 25	147	0.15	< 0.0001	0.821	b	0.706	0.955	-1.109				
					> 25	391	0.13	< 0.0001	1.022	a	0.932	1.121	-1.533				
6(a)	$R_{\text{dark,a}}^{25}$	Leaf [N] <sub>a</sub>	All bar C4H	0.5081	BIT	794	0.10	< 0.0001	1.134		1.061	1.211	-0.296	< 0.0001	-0.300	a	< 0.0001
					C3H	74	0.30	< 0.0001	1.169		0.961	1.421	-0.071				
					NIT	30	0.32	0.0010	1.005		0.735	1.375	-0.287				
					S	132	0.26	< 0.0001	1.257		1.084	1.458	-0.215				
6(d)	$R_{\text{dark,m}}^{25}$	Leaf [N] <sub>m</sub>	All bar C4H	0.0093	BIT	805	0.11	< 0.0001	1.423	a	1.333	1.519	-0.781	< 0.0001	-0.180	b	< 0.0001
					C3H	74	0.60	< 0.0001	1.598	a	1.379	1.852	-0.818				
					NIT	39	0.09	<b>0.0576</b>	2.354		1.723	3.217	-1.763				
					S	132	0.43	< 0.0001	1.383	a	1.213	1.576	-0.579				
6(b)	$R_{\text{dark,a}}^{25}$	Leaf [N] <sub>a</sub>	All bar C4H	0.0512	< 10	47	0.14	0.0109	1.224	a, b	0.929	1.613	-0.008	< 0.0001	0.025	a	< 0.0001
					10 to 15	37	0.15	0.0170	1.700	a	1.245	2.320	-0.399				
					15 to 20	92	0.25	< 0.0001	1.170	b	0.976	1.401	-0.198				
					20 to 25	345	0.29	< 0.0001	1.141	b	1.043	1.248	-0.256				
					> 25	509	0.04	< 0.0001	1.056	b	0.969	1.150	-0.301				
6(e)	$R_{\text{dark,m}}^{25}$	Leaf [N] <sub>m</sub>	All bar C4H	0.0005	< 10	47	0.60	< 0.0001	1.821	a	1.508	2.198	-1.056	< 0.0001	-0.316	d	< 0.0001
					10 to 15	37	0.44	< 0.0001	2.040	a	1.583	2.629	-1.415				
					15 to 20	108	0.44	< 0.0001	1.695	a, b	1.468	1.956	-0.941				

					20 to 25	350	0.36	< 0.0001	1.451	b, c	1.334	1.579	-0.772
					> 25	508	0.06	< 0.0001	1.333	c	1.225	1.451	-0.695
6 (c)	$R_{\text{dark,a}}^{25}$	Leaf [N] <sub>a</sub>	BIT only	0.0004	< 10	4	0.90	<b>0.0537</b>	10.773		4.514	25.707	-3.357
					10 to 15	34	0.10	<b>0.0714</b>	1.680		1.201	2.350	-0.389
					15 to 20	76	0.20	< 0.0001	1.320	a	1.075	1.621	-0.214
					20 to 25	186	0.28	< 0.0001	1.002	b	0.886	1.133	-0.278
					> 25	494	0.03	< 0.0001	1.050	b	0.963	1.146	-0.301
6 (f)	$R_{\text{dark,m}}^{25}$	Leaf [N] <sub>m</sub>	BIT only	0.0041	< 10	4	0.97	0.0161	2.677	a	1.591	4.503	-2.491
					10 to 15	34	0.38	0.0001	2.140	a	1.616	2.833	-1.547
					15 to 20	85	0.44	< 0.0001	1.586	a, b	1.347	1.868	-0.799
					20 to 25	189	0.26	< 0.0001	1.479	b	1.307	1.674	-0.881
					> 25	493	0.05	< 0.0001	1.346	b	1.235	1.467	-0.713

**Table S4. Comparison of linear mixed-effects models with area-based leaf respiration at 25°C ( $R_{\text{dark,a}}^{25}$ ;  $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ) as the response variable (each showing fixed and random effects), with input data restricted to site:species means for which all potential fixed effect parameters were available.** Several model frameworks are outlined (a ‘best predictor model, followed by a null model using PFTs only as fixed factors, then models relevant to different model frameworks, here called ‘ESM’ frameworks), each containing different combinations of fixed effect parameter values (ESM#1-4; for details of each framework, see below). For the fixed effects sub-table, parameter values, s.e. and *t*-values given for the continuous explanatory variables; explanatory variables (all centred on their means) are: (1) plant functional types (PFT), according to *JULES* (Clark *et al.*, 2011): BIT (broad-leaved tree), C3H (C<sub>3</sub> metabolism herbs/grasses), NIT (needle-leaved trees), and S (shrubs); (2) area-based or mass-based leaf nitrogen [ $N_a$  ( $\text{g m}^{-2}$ ) or  $N_m$  ( $\text{mg g}^{-1}$ ), respectively] area-based phosphorus ( $P_a$ ;  $\text{g m}^{-2}$ ) concentrations, area-based Rubisco CO<sub>2</sub> fixation capacity at 25°C ( $V_{\text{cmax,a}}^{25}$ ;  $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ), and mean temperature of the warmest quarter (TWQ; °C) (Hijmans *et al.*, 2005). The PFT-BIT values (first row) are based on the assumption that other variables were at their global mean values. In the ‘best’ model (i.e. same as that shown in Table 5 and Figure 9 in the main text), all available and relevant parameters were included in model selection (PFTs,  $V_{\text{cmax,a}}^{25}$ ,  $N_a$ ,  $P_a$ , TWQ, precipitation of the warmest quarter (PWQ) and aridity index (AI). The null model provides a model where fixed effect factor is limited to PFTs. For ESM#1, the model was limited to the following source fixed effect parameters: PFT,  $N_m$  and  $V_{\text{cmax,a}}^{25}$  and TWQ. Here, our decision to include mass-based N was based on the fact that mass-based N is a predictive trait used in *JULES*, according to Schulze *et al.* (1994). For ESM#2, source fixed effect parameters were the same as for ESM#1, but without  $V_{\text{cmax,a}}^{25}$ . For ESM#3, input fixed effect parameters were: PFT,  $N_a$  and TWQ, while for ESM#4, they were PFT,  $V_{\text{cmax,a}}^{25}$  and TWQ. In the random effect sub-table, the intercept was allowed to vary among species, families and sites; residual errors shown are within species, families, sites and investigators. Finally, predictive equations are shown that enable  $R_{\text{dark,a}}^{25}$  to be predicted based on inputs according to the six models (see next page).

Fixed effect	Best predictor model			Null model (PFT only)			ESM #1			ESM #2			ESM #3			ESM #4		
	Estimate	S.E.	t value	Estimate	S.E.	t value	Estimate	S.E.	t value	Estimate	S.E.	t value	Estimate	S.E.	t value	Estimate	S.E.	t value
PFT_JULES_BIT (if other variables were at global mean)	1.2636	0.033	38.551	1.3805	0.046	29.750	1.2704	0.011	119.349	1.3000	0.012	105.939	1.2855	0.011	117.099	1.2618	0.011	118.611
PFT_JULES_C3H	0.4708	0.141	3.348	0.5099	0.160	3.185	0.3591	0.027	13.135	0.3642	0.030	12.232	0.4395	0.028	15.657	0.4120	0.027	15.176
PFT_JULES_NIT	-0.3595	0.150	-2.392	-0.0558	0.179	-0.311	0.0657	0.033	1.989	-0.0272	0.036	-0.748	-0.2566	0.036	-7.175	0.0259	0.033	0.782
PFT_JULES_S	0.3290	0.064	5.163	0.3460	0.071	4.867	0.3028	0.015	20.290	0.2873	0.016	17.704	0.3188	0.015	20.655	0.3141	0.015	21.009
Leaf [N] (units vary with model, see note below)	0.0728	0.018	4.124				0.0075	0.001	12.574	0.0104	0.001	16.077	0.2061	0.004	46.314			
Leaf_Pa	0.0015	0.000	7.389															
Vcmax_a_25	0.0095	0.001	15.241				0.0114	0.000	58.237							0.0116	0.000	59.229
MeanT_Warmest.Qtr	-0.0358	0.006	-5.658				-0.0338	0.002	-17.949	-0.0389	0.002	-17.983	-0.0402	0.002	-21.055	-0.0334	0.002	-17.766
PFT_JULES_C3H : Leaf_Na	0.3394	0.069	4.892															
PFT_JULES_NIT : Leaf_Na	0.0762	0.146	0.523															
PFT_JULES_S : Leaf_Na	0.0687	0.053	1.295															
Random effect	No. levels per group	Variance compone	% of total	Variance compone	% of total	No. levels per group	Variance compone	% of total	Variance compone	% of total	Variance compone	% of total	Variance compone	% of total	Variance compone	% of total	Variance compone	% of total
Intercept variance: Among species	531	0.009	7.1%	0.023	11.5%	655	0.000	0.0%	0.000	0.0%	0.000	1.9%	0.000	0.0%	0.000	0.0%	0.000	0.0%
Intercept variance: Among families	100	0.002	1.4%	0.004	2.1%	114	0.000	2.4%	0.001	3.0%	0.001	4.5%	0.001	2.6%	0.000	2.6%	0.000	2.6%
Intercept variance: Among sites	49	0.031	23.4%	0.073	36.2%	64	0.005	32.8%	0.006	36.6%	0.005	31.1%	0.005	32.5%	0.005	32.5%	0.005	32.5%
Residual (within species, families and sites plus real error)		0.091	68.2%	0.102	50.2%		0.009	64.8%	0.010	60.5%	0.009	62.4%	0.009	62.4%	0.009	64.9%	0.009	64.9%
		0.133	100.0%	0.202	100.0%		0.014	100.0%	0.017	100.0%	0.015	100.0%	0.014	100.0%	0.014	100.0%	0.014	100.0%
Likelihood ratio test		-595.2		-681.9			-633.7		-689.7		-653.1		-631.9					
Akaike (AIC)		1.220		1.380			1.289		1.399		1.326		1.284					
Bayesian (BIC)		1.288		1.416			1.341		1.446		1.373		1.331					
REML criterion at convergence		1.190		1.364			1.267		1.379		1.306		1.264					
Number of observations (Site:Sp averaged)	667					802												

Continuous explanatory variables HAVE been centred on their means.  
Leaf N is included on an AREA basis in models: Best, Null and ESM 3. And on a MASS basis in ESM 1 and 2.  
The Best and Null models are run on a smaller subset of data given the requirement for Leaf P values.

## Table S4. Continued

### Best predictor model (from Table 6 in the main text)

Broad-leaved trees:  $R_{\text{dark},a}^{25} = 1.236 + (0.0728*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax},a}^{25}) - (0.0358*TWQ)$

C<sub>3</sub> herbs/grasses:  $R_{\text{dark},a}^{25} = 1.7344 + (0.4122*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax},a}^{25}) - (0.0358*TWQ)$

Needle-leaved trees:  $R_{\text{dark},a}^{25} = 0.9041 + (0.1489*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax},a}^{25}) - (0.0358*TWQ)$

Shrubs:  $R_{\text{dark},a}^{25} = 1.5926 + (0.1415*[N]_a) + (0.015*[P]_a) + (0.0095*V_{\text{cmax},a}^{25}) - (0.0358*TWQ)$

### Null model (PFT only) (from Table 6 in the main text)

Broad-leaved trees:  $R_{\text{dark},a}^{25} = 1.3805$

C<sub>3</sub> herbs/grasses:  $R_{\text{dark},a}^{25} = 1.8904$

Needle-leaved trees:  $R_{\text{dark},a}^{25} = 1.3247$

Shrubs:  $R_{\text{dark},a}^{25} = 1.7265$

### ESM#1

Broad-leaved trees:  $R_{\text{dark},a}^{25} = 1.2704 + (0.0075*[N]_m) + (0.0114*V_{\text{cmax},a}^{25}) - (0.0338*TWQ)$

C<sub>3</sub> herbs/grasses:  $R_{\text{dark},a}^{25} = 1.6295 + (0.0075*[N]_m) + (0.0114*V_{\text{cmax},a}^{25}) - (0.0338*TWQ)$

Needle-leaved trees:  $R_{\text{dark},a}^{25} = 1.3361 + (0.0075*[N]_m) + (0.0114*V_{\text{cmax},a}^{25}) - (0.0338*TWQ)$

Shrubs:  $R_{\text{dark},a}^{25} = 1.5732 + (0.0075*[N]_m) + (0.0114*V_{\text{cmax},a}^{25}) - (0.0338*TWQ)$

### ESM#2

Broad-leaved trees:  $R_{\text{dark},a}^{25} = 1.300 + (0.0104*[N]_m) - (0.0389*TWQ)$

C<sub>3</sub> herbs/grasses:  $R_{\text{dark},a}^{25} = 1.66642 + (0.0104*[N]_m) - (0.0389*TWQ)$

Needle-leaved trees:  $R_{\text{dark},a}^{25} = 1.2728 + (0.0104*[N]_m) - (0.0389*TWQ)$

Shrubs:  $R_{\text{dark},a}^{25} = 1.5875 + (0.0104*[N]_m) - (0.0389*TWQ)$

### ESM#3

Broad-leaved trees:  $R_{\text{dark},a}^{25} = 1.2855 + (0.2061*[N]_a) - (0.0402*TWQ)$

C<sub>3</sub> herbs/grasses:  $R_{\text{dark},a}^{25} = 1.7250 + (0.2061*[N]_a) - (0.0402*TWQ)$

Needle-leaved trees:  $R_{\text{dark},a}^{25} = 1.0290 + (0.2061*[N]_a) - (0.0402*TWQ)$

Shrubs:  $R_{\text{dark},a}^{25} = 1.6043 + (0.2061*[N]_a) - (0.0402*TWQ)$

### ESM#4

Broad-leaved trees:  $R_{\text{dark},a}^{25} = 1.2818 + (0.0116 * V_{\text{cmax},a}^{25}) - (0.0334*TWQ)$

C<sub>3</sub> herbs/grasses:  $R_{\text{dark},a}^{25} = 1.6737 + (0.0116 * V_{\text{cmax},a}^{25}) - (0.0334*TWQ)$

Needle-leaved trees:  $R_{\text{dark},a}^{25} = 1.2877 + (0.0116 * V_{\text{cmax},a}^{25}) - (0.0334*TWQ)$

Shrubs:  $R_{\text{dark},a}^{25} = 1.5758 + (0.0116 * V_{\text{cmax},a}^{25}) - (0.0334*TWQ)$

**Table S5. Comparison of linear mixed-effects models using different plant functional types (PFT) classifications, with leaf respiration at 25°C ( $R_{\text{dark}}^{25}$ ) as the response variable.** Two models are shown: (A) using area-based leaf respiration at 25°C ( $R_{\text{dark,a}}^{25}$ ;  $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ); and, (B) mass-based leaf respiration at 25°C ( $R_{\text{dark,m}}^{25}$ ;  $\text{nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$ ). For (A) and (B), two model frameworks are outlined (variants of ESM#3 model shown in Table S4, but with a larger number of observations reflecting the abundance of  $[\text{N}]_{\text{a}}$  ( $\text{g m}^{-2}$ ) and  $[\text{N}]_{\text{m}}$  ( $\text{mg g}^{-1}$ ) data), differing in the plant functional types (PFT) used: *JULES*(Clark *et al.*, 2011): BIT (broad-leaved tree), C3H ( $\text{C}_3$  metabolism herbs/grasses), NIT (needle-leaved trees), and S (shrubs); and, *LPJ* (Sitch *et al.*, 2003): BorDcBl, boreal deciduous broad-leaved tree/shrub; BorDcNI, boreal deciduous needle-leaved tree/shrub; BorEvNI, boreal evergreen needle-leaved tree/shrub; TmpDcBl, temperate deciduous broad-leaved tree/shrub; TmpEvBl, temperate evergreen broad-leaved tree/shrub; TmpEvNI, temperate evergreen needle-leaved tree/shrub; TmpH, temperate herb/grass; TrpDcBl, tropical deciduous broad-leaved tree/shrub; TrpEvBl, tropical evergreen broad-leaved tree/shrub; TrpH, tropical herb/grass. For the fixed effects sub-tables, parameter values, s.e. and *t*-values given for the continuous explanatory variables; explanatory variables (all centred on their means) are: PFTs; area or mass-based leaf nitrogen ( $N_{\text{a}}$  and  $N_{\text{m}}$ , respectively) and mean temperature of the warmest quarter (TWQ) (Hijmans *et al.*, 2005). For *JULES*, the PFT-BIT values (first row) are based on the assumption that other variables were at their global mean values. Similarly, for *LPJ*, the PFT-BorDcBl (first row) are based on the assumption that other variables were at their global mean values. In the random effect sub-table, the intercept was allowed to vary among species, families and sites; residual errors shown are within species, families, sites and investigators.

(A) area-based

JULES				LPJ			
Fixed effect	Estimate	S.E.	t value	Fixed effect	Estimate	S.E.	t value
PFT_JULES_BIT (if other variables were at global mean)	1.1911	0.034	35.041	PFT_LPJ_BorDcBl	1.3667	0.152	8.982
PFT_JULES_C3H	0.3930	0.069	5.709	PFT_LPJ_BorDcNI	0.2756	0.166	1.659
PFT_JULES_NIT	0.1392	0.091	1.536	PFT_LPJ_BorEvNI	-0.5574	0.234	-2.382
PFT_JULES_S	0.3298	0.045	7.298	PFT_LPJ_TmpDcBl	-0.1969	0.152	-1.292
				PFT_LPJ_TmpEvBl	0.0018	0.149	0.012
				PFT_LPJ_TmpEvNI	-0.0581	0.177	-0.329
				PFT_LPJ_TmpH	0.1723	0.151	1.141
				PFT_LPJ_TrpDcBl	-0.1469	0.167	-0.881
				PFT_LPJ_TrpEvBl	-0.1785	0.162	-1.104
				PFT_LPJ_TrpH	0.0177	0.267	0.066
Leaf $[\text{N}]_{\text{area}}$	0.2589	0.014	18.973	Leaf $[\text{N}]_{\text{area}}$	0.2592	0.014	18.407
MeanT_Warmest.Qtr	-0.0396	0.006	-6.381	MeanT_Warmest.Qtr	-0.0351	0.007	-4.743

Random effect	Variance component		Variance component	
		% of total		% of total
Intercept variance: Among species	0.000	0.0%	0.000	0.0%
Intercept variance: Among families	0.008	3.4%	0.007	2.8%
Intercept variance: Among sites	0.056	24.1%	0.061	25.7%
Residual (within species, families and sites plus real error)	0.167	72.5%	0.170	71.5%
	0.230	100.0%	0.237	100.0%

logLikelihood	-1,011	-1,022
Akaike (AIC)	2,041	2,075
Bayesian (BIC)	2,091	2,154
REML criterion at convergence	2,021	2,043

Number of obs: 1025  
Groups: Species, 833; Family, 129; Site, 81

Continuous explanatory variables HAVE been centred on their means

(B) mass-based

JULES				LPJ			
Fixed effect	Estimate	S.E.	t value	Fixed effect	Estimate	S.E.	t value
PFT_JULES_BIT (if other variables were at global mean)	11.0413	0.068	161.680	PFT_LPJ_BorDcBl	14.7990	0.254	58.233
PFT_JULES_C3H	4.1153	0.125	32.816	PFT_LPJ_BorDcNI	0.7859	0.409	1.921
PFT_JULES_NIT	-2.0938	0.156	-13.408	PFT_LPJ_BorEvNI	-7.7081	0.303	-25.481
PFT_JULES_S	1.6953	0.073	23.248	PFT_LPJ_TmpDcBl	-2.8388	0.258	-11.000
				PFT_LPJ_TmpEvBl	-3.5225	0.250	-14.069
				PFT_LPJ_TmpEvNI	-5.7636	0.293	-19.703
				PFT_LPJ_TmpH	0.2726	0.257	1.062
				PFT_LPJ_TrpDcBl	-3.5374	0.288	-12.268
				PFT_LPJ_TrpEvBl	-3.8196	0.275	-13.872
				PFT_LPJ_TrpH	-2.3686	0.476	-4.974
Leaf $[\text{N}]_{\text{mass}}$	0.4586	0.003	135.520	Leaf $[\text{N}]_{\text{mass}}$	0.4567	0.003	133.980
MeanT_Warmest.Qtr	-0.3842	0.013	-30.559	MeanT_Warmest.Qtr	-0.3353	0.015	-23.104

Random effect	Variance component		Variance component	
		% of total		% of total
Intercept variance: Among species	0.000	0.0%	0.000	0.0%
Intercept variance: Among families	0.036	8.9%	0.033	8.2%
Intercept variance: Among sites	0.258	64.3%	0.262	65.0%
Residual (within species, families and sites plus real error)	0.108	26.9%	0.108	26.8%
	0.401	100.0%	0.403	100.0%

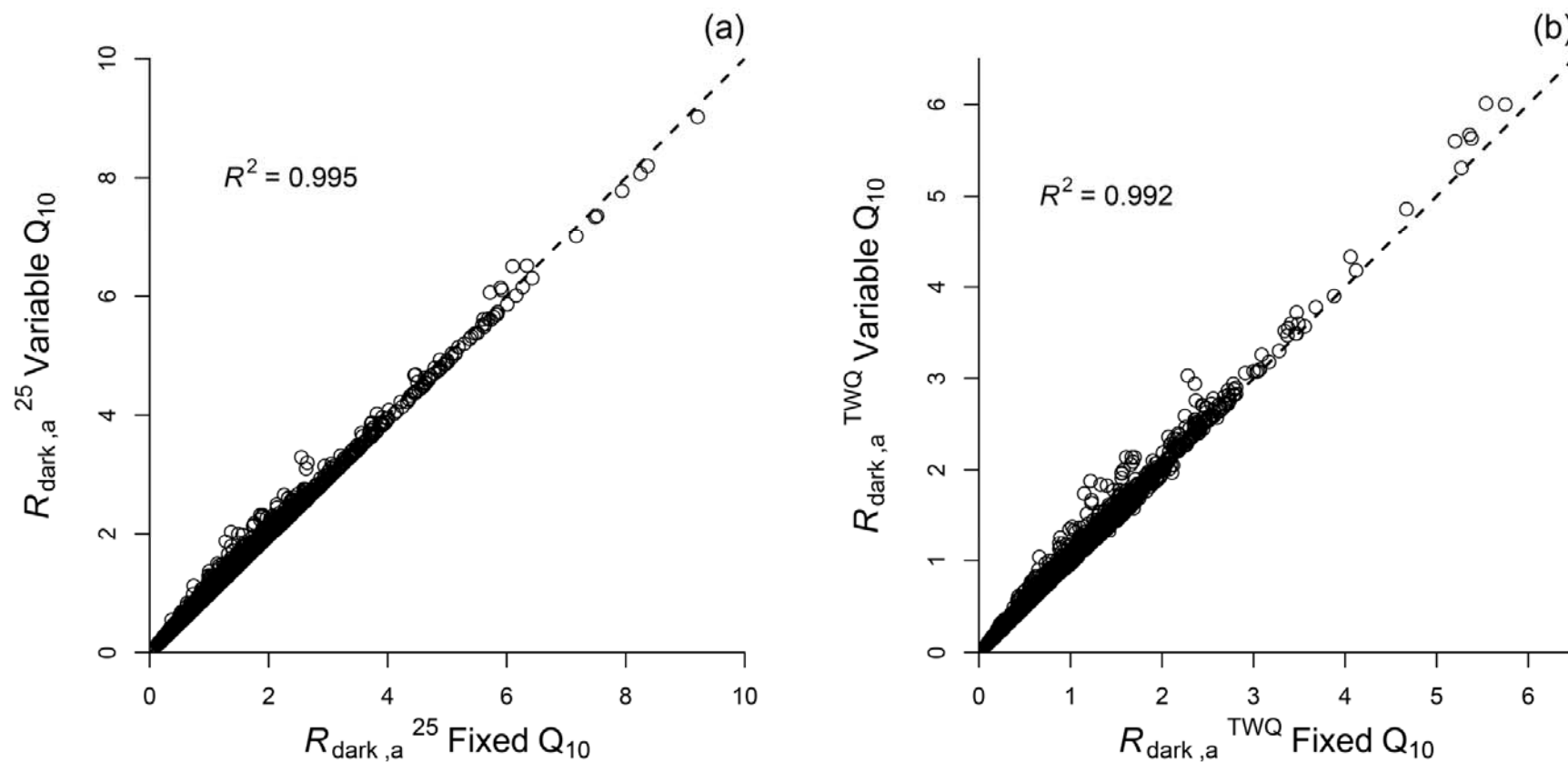
logLikelihood	-2,655	-2,644
Akaike (AIC)	5,330	5,319
Bayesian (BIC)	5,379	5,398
REML criterion at convergence	5,310	5,287

Response variable is Rdarkm\_25C\_varQ10

Continuous explanatory variables HAVE been centred on their means

Number of obs: 1045  
Groups: Species, 833; Family, 129; Site, 87

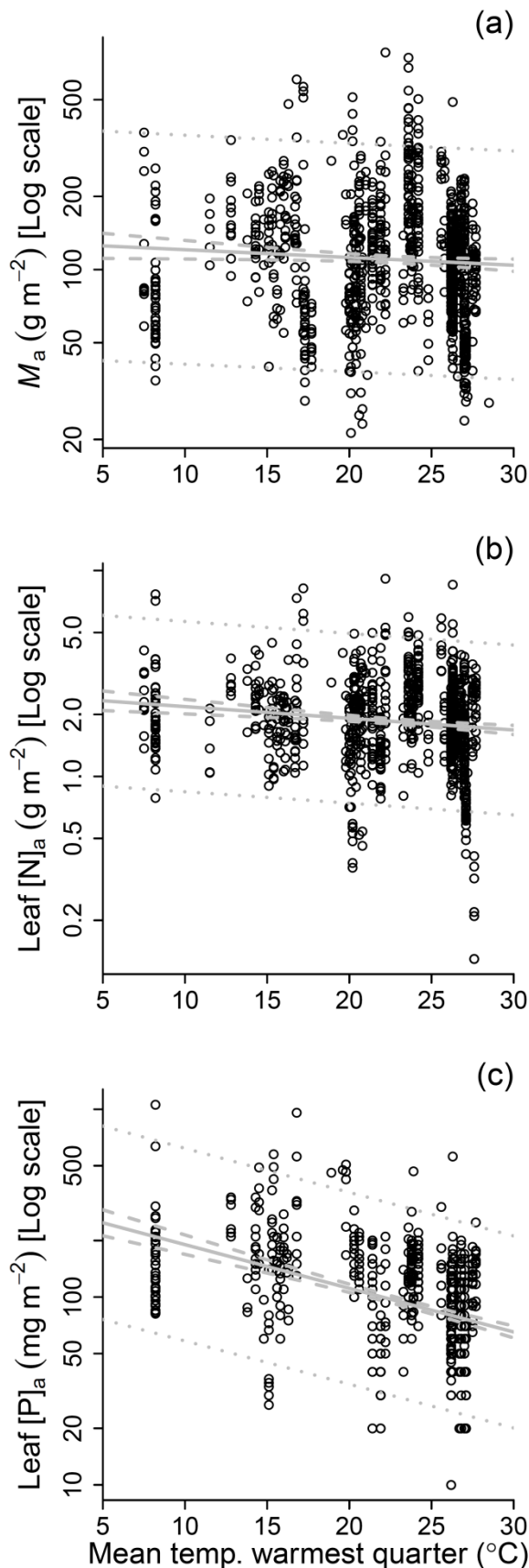
**Figure S1. Comparison of area-based rates of leaf respiration in darkness ( $R_{\text{dark}}$ ) at a common leaf temperature of 25°C, calculated assuming either a fixed  $Q_{10}$  of 2.23 (Atkin *et al.*, 2005) (using Eqn 1 in the main text) or assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) (using Eqn 2 in the main text).  $R_{\text{dark},a}^{25}$  and  $R_{\text{dark},a}^{\text{TWQ}}$ , predicted area-based  $R_{\text{dark}}$  rates ( $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ) at 25°C, and TWQ (mean  $T$  of the warmest quarter), respectively. Values at the TWQ of each replicate were calculated using climate data from the *WorldClim* data base (Hijmans *et al.*, 2005). Data shown are for individual observational rows in the global respiration database.**



**Figure S2. Relationships between leaf structural and chemical composition traits, and mean daily temperature of the warmest quarter (TWQ).**

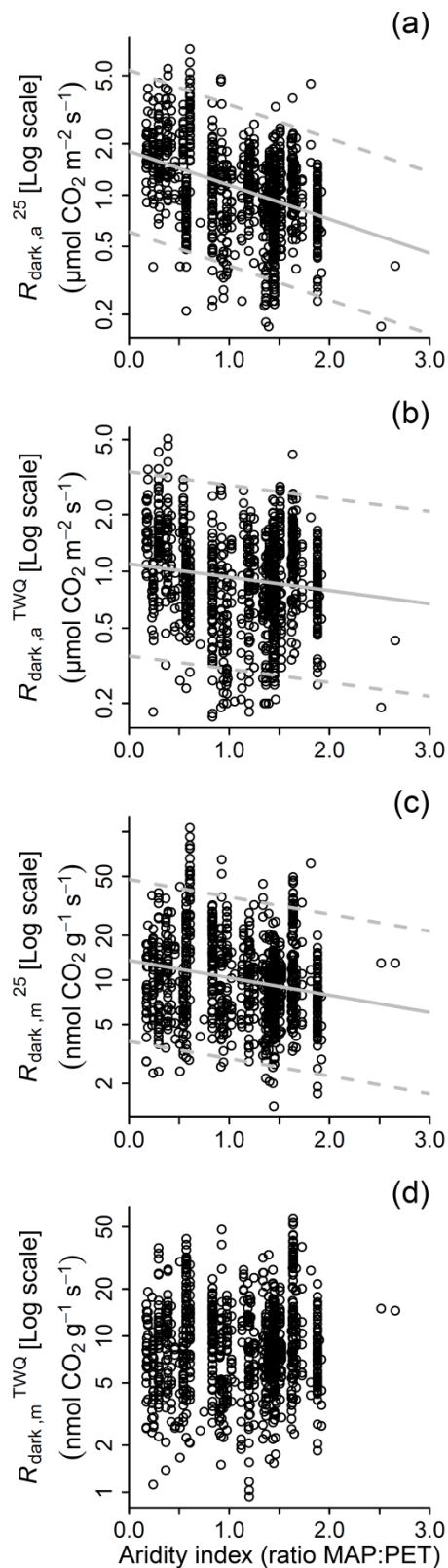
Values shown are averages for unique site:species combinations in the global *GlobResp* database. Traits shown are: (a)  $M_a$ , leaf mass per unit leaf area; (b)  $[N]_a$ , area-based leaf nitrogen concentration; and (c)  $[P]_a$ , area-based leaf phosphorous concentration. TWQ at each site were obtained using site information and the *WorldClim* data base (Hijmans *et al.*, 2005). Solid grey line in each plot shows regression lines where the relationships were significant (with 95% confidence intervals shown as dashed line around the predicted relationship; the dotted lines show the prediction intervals (two-times the standard deviation) around the predicted relationship).

While the negative  $M_a \leftrightarrow TWQ$  (Fig. S2a) and  $[N]_a \leftrightarrow TWQ$  (Fig. 4b) relationships were both significant ( $M_a$ :  $p < 0.05$ ,  $n = 1092$ ;  $[N]_a$ :  $p < 0.0001$ ,  $n = 1029$ ), in neither case were the associations strong ( $M_a$ : Pearson's correlation ( $r$ ) = -0.067,  $r^2 = 0.004$ ;  $[N]_a$ :  $r = -0.134$ ,  $r^2 = 0.018$ ). By contrast, the negative  $[P]_a \leftrightarrow TWQ$  relationship (Fig. 4c) was more marked ( $p < 0.0001$ ,  $n = 728$ ,  $r = -0.418$ ,  $r^2 = 0.174$ ), with  $[P]_a$  being highest at the coldest sites.



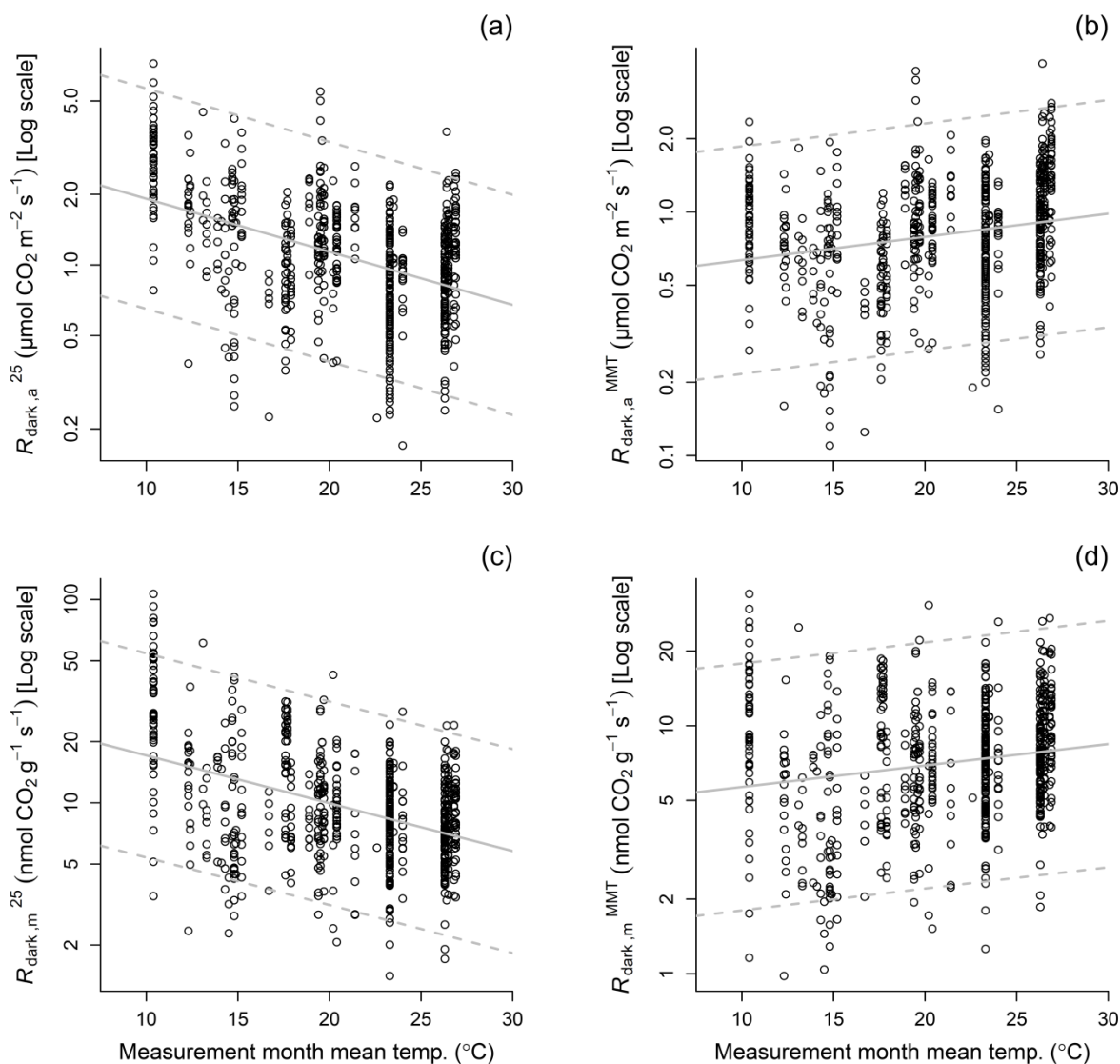


**Figure S3. Site-species mean values leaf  $R_{\text{dark}}$  ( $\log_{10}$  scale) relationships with aridity index (AI), excluding data from the exceptionally high-rainfall, Frans Josef Glacier (FJG) site in NZ.** Traits shown are:  $R_{\text{dark},a}^{25}$ , (a) and  $R_{\text{dark},a}^{\text{TWQ}}$  (b), predicted area-based  $R_{\text{dark}}$  rates at 25°C and TWQ, respectively;  $R_{\text{dark},m}^{25}$  (c) and  $R_{\text{dark},m}^{\text{TWQ}}$  (d), predicted mass-based  $R_{\text{dark}}$  rates at 25°C and TWQ, respectively. Values at 25°C and TWQ were calculated assuming a temperature-dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). Values at the TWQ of each replicate were calculated using climate/location data from the *WorldClim* data base (Hijmans *et al.*, 2005). Aridity index calculated as the ratio of mean annual precipitation (MAP) to mean annual potential evapotranspiration (PET) (UNEP, 1997). Solid lines in each plot show regression lines where the relationships were significant; dashed lines show the prediction intervals (two-times the standard deviation) around the predicted relationship. See Figure 4 for the same figure where data from FJG were included.

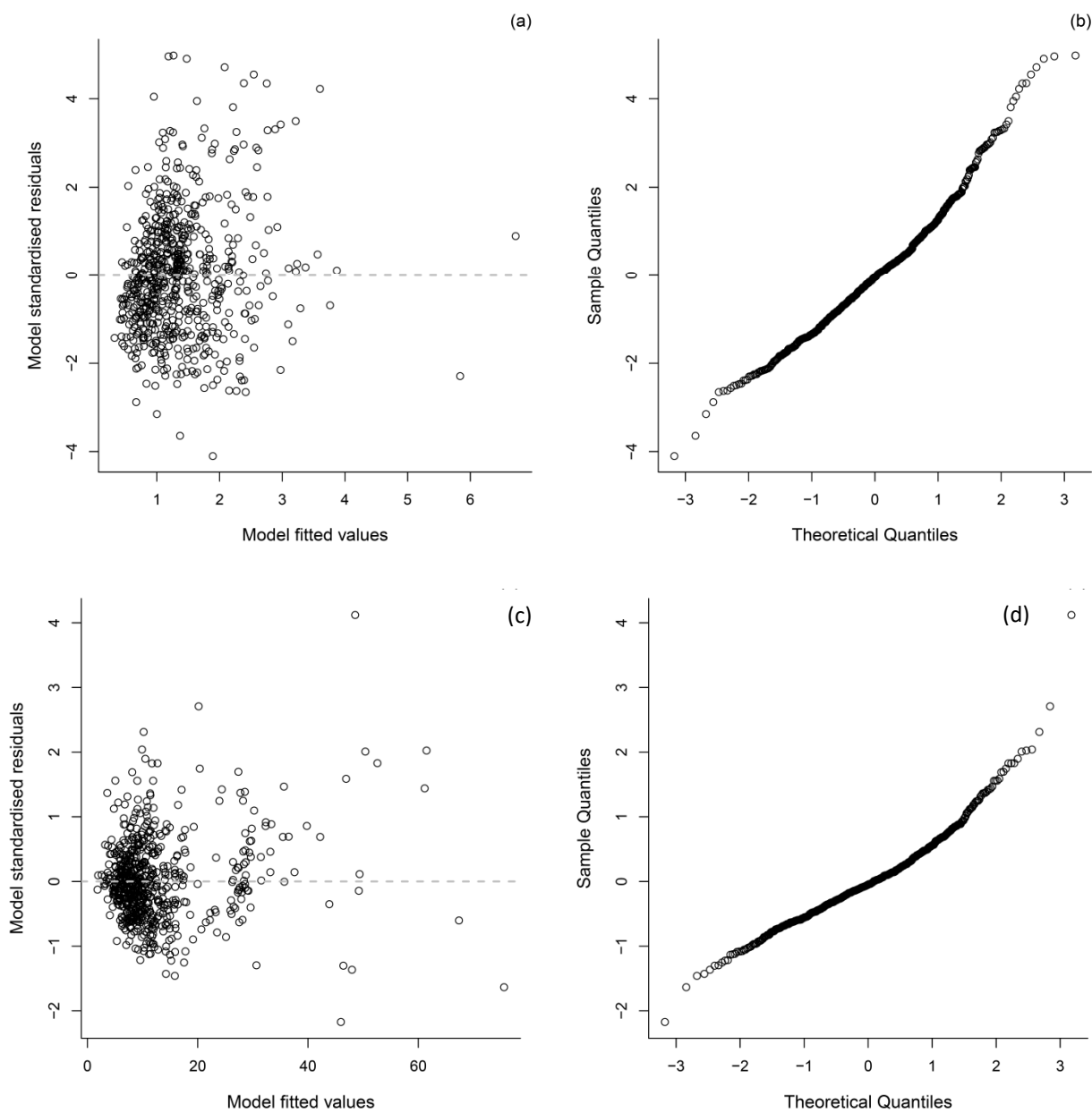


**Figure S4. Relationships between leaf  $R_{\text{dark}}$  ( $\log_{10}$  scale) and measuring month mean daily temperature (MMT) for those sites where the month of measurement was known.** Values shown are averages for unique site:species combinations, using previously unpublished data (Supporting Information Table S1). Traits shown are: (a)  $R_{\text{dark},a}^{25}$ , predicted area-based  $R_{\text{dark}}$  at 25°C; (b)  $R_{\text{dark},a}^{\text{MMT}}$ , predicted area-based  $R_{\text{dark}}$  at MMT; (c)  $R_{\text{dark},m}^{25}$ , mass-based  $R_{\text{dark}}$  at 25°C; (d)  $R_{\text{dark},m}^{\text{MMT}}$ , mass-based  $R_{\text{dark}}$  at MMT. Values at 25°C and MMT were calculated assuming a  $T$ -dependent  $Q_{10}$  (Tjoelker *et al.*, 2001) and equation 7 described in Atkin *et al.* (2005). Values at the MMT of each replicate were calculated using climate/location data from the *WorldClim* data base (Hijmans *et al.*, 2005). Solid lines in each plot show regression lines where the relationships were significant; dashed lines show the prediction intervals (two-times the standard deviation) around the predicted relationship.

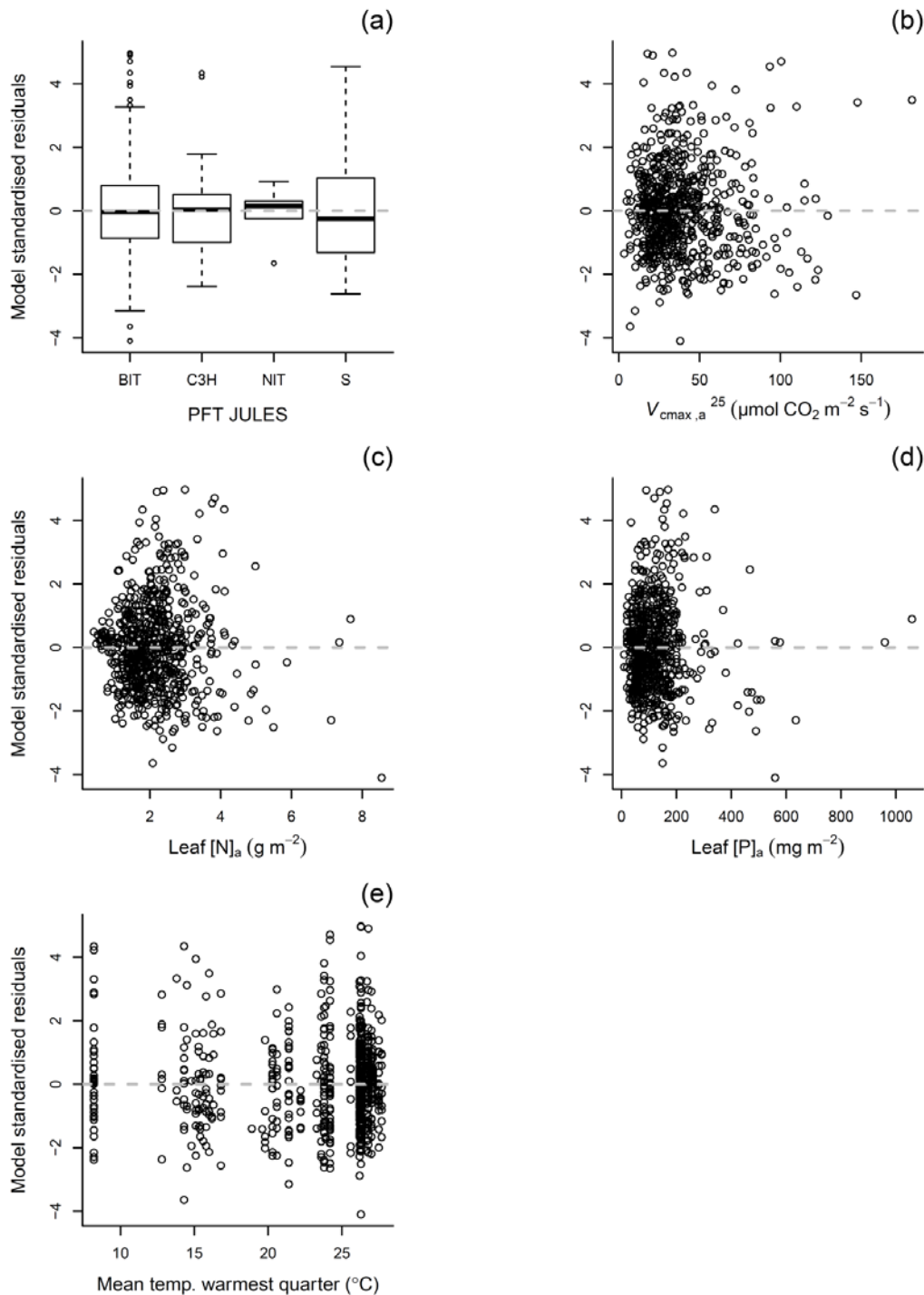
For  $R_{\text{dark},a}^{25}$ , the negative relationship with MMT was significant ( $p < 0.0001$ ,  $n = 677$ ,  $r^2 = 0.192$ ;  $\log_{10} R_{\text{dark},a}^{25} = 0.509 - 0.023 * \text{MMT}$ ) (Fig. S4a). Similarly, the  $R_{\text{dark},a}^{\text{MMT}} \leftrightarrow \text{MMT}$  association (Fig. S4b) was significant ( $p < 0.0001$ ,  $n = 677$ ,  $r^2 = 0.041$ ;  $\log_{10} R_{\text{dark},a}^{\text{MMT}} = -0.293 + 0.0095 * \text{MMT}$ ), as were the  $R_{\text{dark},m}^{25} \leftrightarrow \text{MMT}$  ( $p < 0.0001$ ,  $n = 667$ ,  $r^2 = 0.184$ ;  $\log_{10} R_{\text{dark},m}^{25} = 1.468 - 0.023 * \text{MMT}$ ) and  $R_{\text{dark},m}^{\text{MMT}} \leftrightarrow \text{MMT}$  ( $p < 0.0001$ ,  $n = 667$ ,  $r^2 = 0.030$ ;  $\log_{10} R_{\text{dark},m}^{\text{MMT}} = 0.666 + 0.009 * \text{MMT}$ ) relationships (Fig. S4c,d).



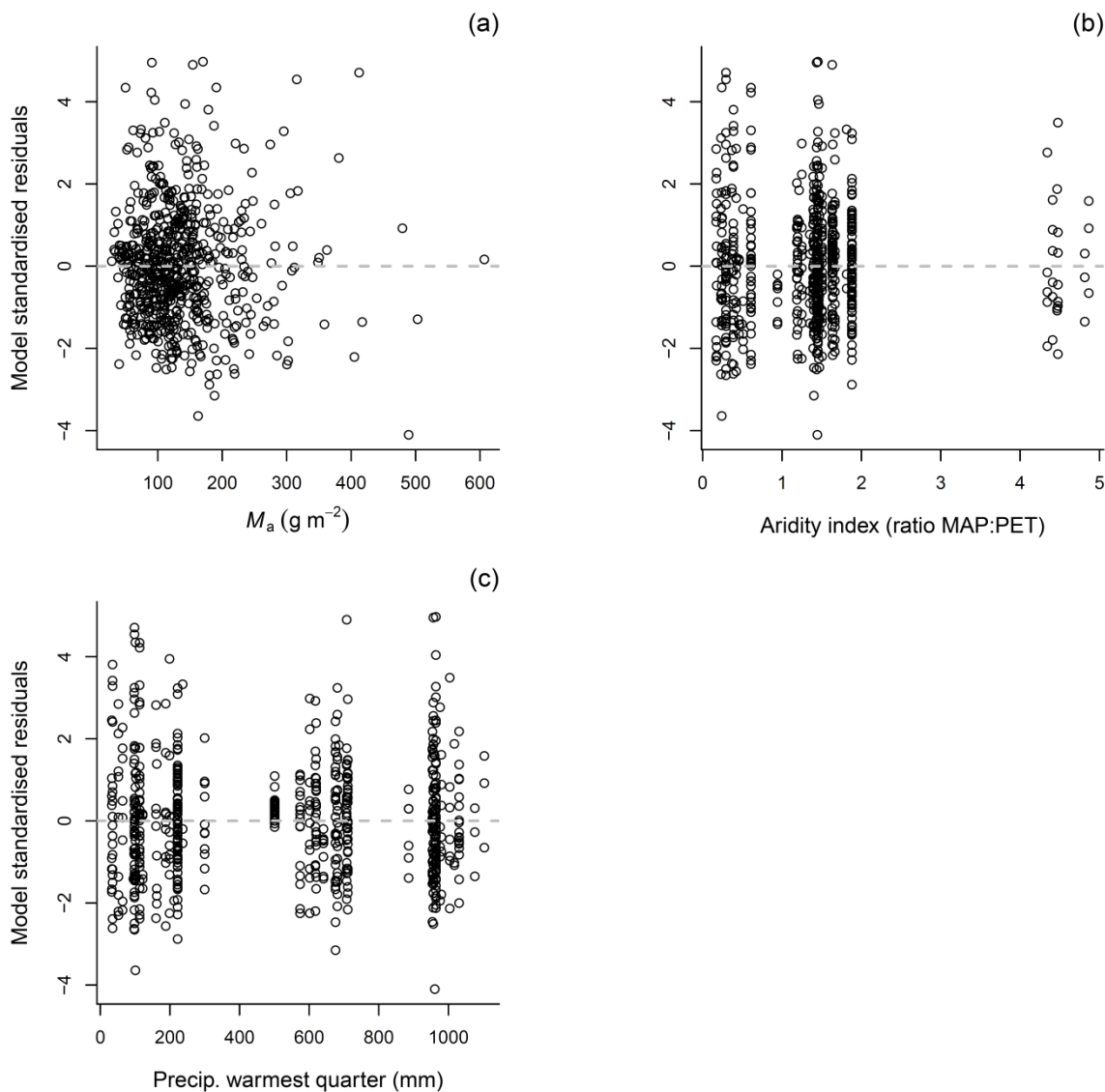
**Figure S5. Testing key assumptions for area- and mass-based mixed effects models –heterogeneity and normality.** See Table 5 in the main text for details on the models. The upper panel [(a) and (b)] refer to the model based on area-based values, while the lower panel [(c) and (d)] refers to the mass-based model.



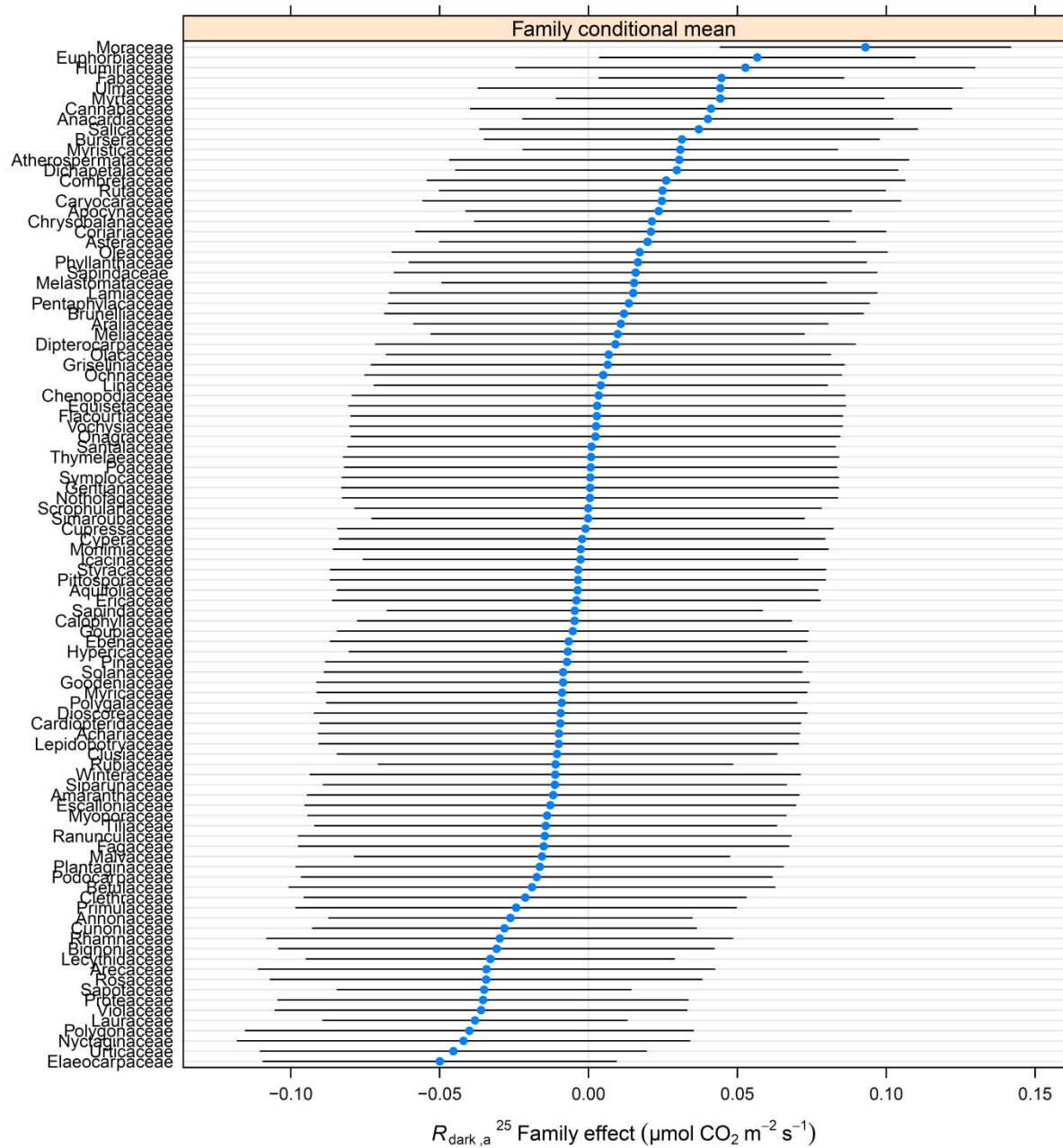
**Figure S6. Model validation graphs for the area-based mixed effects model.** Shown are standardised residuals plotted against fitted values for each of the continuous explanatory factors and variables used in the model's fixed components: (a) plant functional types (PFT) categorised according to *JULES* (BIT, broadleaved trees; C3H, C<sub>3</sub> herbs; NIT, needle-leaved trees; S, shrubs); (b) area-based rates of the  $V_{cmax}$  of Rubisco at 25°C ( $V_{cmax,a}^{25}$ ); (c) leaf nitrogen per unit leaf area ( $[N]_a$ ); (d) leaf phosphorus per unit leaf area ( $[P]_a$ ); and, (e) mean temperature of the warmest quarter at each site. See Table 5 in the main text for details on the models. Similar graphs were made for the mass-based model (data not shown). For (a), the central box in each plot shows the interquartile range; the median is shown as the bold line in each box; whiskers extend 1.5 times the interquartile range or the most extreme value, whichever is smaller; any points outside the values are shown as individual points.



**Figure S7. Standardised residuals plotted against fitted values for variables not used in the area-based model's fixed components.** See Table 5 in the main text for details on the models. Similar graphs were made for the mass-based model (data not shown). Plots show residuals against (a) leaf mass per unit leaf area ( $M_a$ ) categorised; (b) aridity index (ratio of mean annual precipitation to potential evapotranspiration); (c) precipitation of the warmest quarter at each site. See Table 5 in the main text for details on the models.



**Figure S8. Dotchart of the area-based mixed model's random intercepts by Family.** Points represent the difference (shown with 95% prediction intervals) for each family in the  $R_{\text{dark},a}^{25}$  response above or below the overall population mean after controlling for the model's fixed terms and site location (Figure S7). See Table 5 in the main text for details on the models. Similar graphs were made for the mass-based model (data not shown).



**Figure S9. Dotchart of the area-based mixed model's random intercepts by site.** Points represent the difference (shown with 95% prediction intervals) for each site in the  $R_{\text{dark},a}^{25}$  response above or below the overall population mean after controlling for the model's fixed terms and phylogenetic structure (Figure S6). See Table 5 in the main text for details on the models. Similar graphs were made for the mass-based model (data not shown)

