

Supplementary Information for

How functional traits influence plant growth and shade tolerance across the life-cycle

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Supporting Information Text

Additional details on the FF16 growth model in the plant package

In this paper we use the FF16 growth model from the package plant (1) , which is accessed via R (2) . A full derivation of the Eqs. from the main text are described is provided in Appendix of [\(1\)](#page-12-1), see also [\(3\)](#page-12-3). Below we repeat some additional details from [\(1\)](#page-12-1), so that readers have a complete picture of the model.

The FF16 physiological model includes default values for all needed parameters (Table [S1\)](#page-5-0). Species are known to vary considerably in many of these parameters, such as ϕ , ρ , ν , and ω ; so by varying parameters one can account for species differences. When altering a parameter in the model, however, one must also consider whether there are trade-offs linking parameters. This is achieved via a hyper-parameterisation function.

Hyper-parameterisation of the FF16 model via traits. plant allows for the hyper-parameterisation of the FF16 physiological model via plant functional traits: this enables simultaneous variation in multiple parameters in accordance with an assumed trade-off. The hyper-parameterisation function used in the current analysis differs slightly from that in [\(1\)](#page-12-1). The functions implemented are as follows.

Seed mass. Effects of the trait seed mass, are naturally embedded in the equation determining fecundity and the initial height of seedlings. In addition, we let the accessory cost per seed be a multiple of seed size, $\alpha_{f3} = \beta_{f1}\omega$, as empirically observed [\(4\)](#page-12-4).

Nitrogen per unit leaf area. Photosynthesis per unit leaf area and respiration rates per unit leaf mass (or area) are assumed to vary with leaf nitrogen per unit area, *ν*. The calculation of respiration rates is already described above. To calculate the average annual photosynthesis for a leaf, we integrate the instantaneous rate per unit leaf area over the annual solar trajectory, using a rectangular-hyperbolic photosynthesis light response curve,

$$
p(\nu, E) = \frac{1}{365d} \int_0^{365d} \frac{Y(t) + A_{\text{max}} - \sqrt{(Y(t) + A_{\text{max}})^2 - 4\beta_{\text{lf}2}Y(t)A_{\text{max}}}}{2\beta_{\text{lf}2}} dt, \tag{1}
$$

where A_{max} is the maximum photosynthetic capacity of the leaf, β_{IF2} is the curvature of the light response curve, $Y(t) = \beta_{\text{IF3}} I(t)$ is the initial yield of the light response curve, with β_{B} being the quantum yield parameter, $I(t) = k_I I_0(t) E$ is the intensity of light on the leaf surface, and $I_0(t)$ is light incident on a surface perpendicular to the sun's rays directly above the canopy at time *t*. The profile of $I_0(t)$ is given by a solar model adapted from (?).

We allow for the maximum photosynthetic capacity of the leaf to vary with leaf nitrogen per unit area, as

$$
A_{\max} = \beta_{\text{lf1}} \left(\frac{\nu}{\nu_0} \right)^{\beta_{\text{lf5}}},\tag{2}
$$

where $β_{\text{lf1}}$ and $β_{\text{lf5}}$ are constants. The relationships is normalized around *ν*₀, the global mean of leaf nitrogen per unit area.

Leaf mass per unit area. The trait leaf mass per unit area, denoted by ϕ , directly influences growth by changing dA_1/dM_a . In addition, we link ϕ to the rate of leaf turnover, based on a widely observed scaling relationship from [\(5\)](#page-12-5) (Fig. [S1\)](#page-7-0),

$$
k_1 = \beta_{k11} \left(\frac{\phi}{\phi_0}\right)^{-\beta_{k12}}.
$$
\n
$$
(3)
$$

This relationship is normalised around ϕ_0 , the global mean of ϕ . This allows us to vary β_{k11} and β_{k2} without displacing the relationship from the observed mean.

We also vary the mass-based leaf respiration rate so that it stays constant per unit leaf area and varies with ϕ and nitrogen per unit leaf area ν , as empirically observed by (5) ,

$$
r_1 = \frac{\beta_{\text{lf4}}\,\nu}{\phi}.\tag{4}
$$

Wood density. The trait wood density, denoted by ρ , directly influences growth by changing dA_1/dM_a . As for ϕ , these relationships are normalized around ρ_0 , the global mean of ρ . By default, β_{k12} and β_{ks2} are set to zero, so these linkages only become present when these parameters are set to something other than their default values.

The rate of sapwood respiration per unit volume is assumed to be constant, so sapwood respiration per unit mass varies as

$$
r_{\rm s} = \frac{\beta_{\rm rs1}}{\rho},\tag{5}
$$

where *β*rs1 is a default rate per volume of sapwood. Similarly, the rate of bark respiration per unit mass varies as

$$
r_{\rm b} = \frac{\beta_{\rm rbl}}{\rho},\tag{6}
$$

with $\beta_{\text{rb1}} = 2\beta_{\text{rs1}}$.

Implementation of hyper-parameterisation function. The hyper-parameterisation function used in this analysis is implemented as follows:

```
function(
```

```
lma_0=0.1978791,
                              B_kl1=0.4565855,
                              B_kl2=1.71,
                              rho_0=608.0,
                              B_dI1=0.01,
                              B_dI2=0.0,
                              B_ks1=0.2,
                              B_ks2=1.25, #0.0,
                              B_rs1=4012.0,
                              B_rb1=2.0*4012.0,
                              B_f1 = 3.0,
                              narea=1.87e-3,
                              narea_0=1.87e-3,
                              B_lf1=5120.738 * 1.87e-3 * 24 * 3600 / 1e+06,
                              B_lf2=0.75,
                              B_lf3=0.04,
                              B_lf4=21000*0.75,
                              B_lf5=0.5,
                              k_I = 0.5,
                              latitude=0) {
assert\_scalar \leftarrow function(x, name = deparse(substitute(x))) {
 if (\text{length}(x) := 1L) {
   stop(sprintf("%s must be a scalar", name), call. = FALSE)
 }
}
assert_scalar(lma_0)
assert_scalar(B_kl1)
assert_scalar(B_kl2)
assert scalar(rho 0)
assert_scalar(B_dI1)
assert_scalar(B_dI2)
assert_scalar(B_ks1)
assert_scalar(B_ks2)
assert_scalar(B_rs1)
assert_scalar(B_rb1)
assert scalar(B_f1)
assert_scalar(narea)
assert scalar(narea 0)
assert_scalar(B_lf1)
assert_scalar(B_lf2)
assert_scalar(B_lf3)
assert_scalar(B_lf4)
assert_scalar(B_lf5)
assert_scalar(k_I)
assert_scalar(latitude)
function(m, s, filter=TRUE) {
  with_default <- function(name, default_value=s[[name]]) {
   rep_len(if (name %in% colnames(m)) m[, name] else default_value,
            nrow(m))
  }
  lma <- with_default("lma")
  rho <- with_default("rho")
  omega <- with_default("omega")
  narea <- with_default("narea", narea)
  ## lma / leaf turnover relationship:
  k_l <- B_kl1 * (lma / lma_0) ^ (-B_kl2)
  ## rho / mortality relationship:
```

```
d_I <- B_dI1 * (rho / rho_0) ^ (-B_dI2)
## rho / wood turnover relationship:
k_s <- B_ks1 * (rho / rho_0) ^ (-B_ks2)
k_b \leftarrow k_s## rho / sapwood respiration relationship:
## Respiration rates are per unit mass, so this next line has the
## effect of holding constant the respiration rate per unit volume.
## So respiration rates per unit mass vary with rho, respiration
## rates per unit volume don't.
r_s <- B_rs1 / rho
# bark respiration follows from sapwood
r b \leftarrow B rb1 / rho## omega / accessory cost relationship
a_f3 \leftarrow B_f1 * \text{omega}## Narea, photosynthesis, respiration
assimilation_rectangular_hyperbolae <- function(I, Amax, theta, QY) {
 x \leftarrow QY * I + Amax(x - sqrt(x^2 - 4 * theta * QY * I * Amax)) / (2 * theta)}
## Photosynthesis [mol CO2 / m2 / yr]
approximate_annual_assimilation <- function(narea, latitude) {
  E \leftarrow \text{seq}(0, 1, \text{ by=0.02})## Only integrate over half year, as solar path is symmetrical
  D \leftarrow \text{seq}(0, 365/2, \text{length.out} = 10000)I <- plant:::PAR_given_solar_angle(plant:::solar_angle(D, latitude = abs(latitude)))
  Amax \leq B_lf1 * (narea/narea_0) \hat{B} B_lf5
  theta \leftarrow B_lf2
  QY \leftarrow B_1f3AA \leftarrow NA * Efor (i in seq_len(length(E))) {
    AA[i] <- 2 * plant:::trapezium(D, assimilation_rectangular_hyperbolae(
                                  k_I * I * E[i], Amax, theta, QY)}
  if(all(diff(AA) < 1E-8)) {
    # line fitting will fail if all have are zero, or potentially same value
    ret \leftarrow c(last(AA), 0)names(ret) <- c("p1","p2")
  } else {
    fit <- nls(AA ~ p1 * E/(p2 + E),
              data-frame(E = E, AA = AA),
               start = list(p1 = 100, p2 = 0.2)ret <- coef(fit)
  }
  ret
}
# This needed in case narea has length zero, in which case trapezium fails
a_p1 \leftarrow a_p2 \leftarrow 0 * narea
## TODO: Remove the 0.5 hardcoded default for k_I here, and deal
## with this more nicely.
if (length(narea) > 0 || k_I != 0.5) {
```

```
i <- match(narea, unique(narea))
       y <- vapply(unique(narea), approximate_annual_assimilation,
                     numeric(2), latitude)
       a_{p1} < -y["p1", i]
       a_p2 \leq y["p2", i]
    }
    ## Respiration rates are per unit mass, so convert to mass-based
    ## rate by dividing with lma
    ## So respiration rates per unit mass vary with lma, while
    ## respiration rates per unit area don't.
    r_l <- B_lf4 * narea / lma
    extra \leftarrow cbind(k_1, # lma
                      d_I, k_s, k_b, r_s, r_b, # rho
                       a_f3, \qquad \qquada_p1, a_p2, # narea
                       r_1) \# lma, narea
    overlap <- intersect(colnames(m), colnames(extra))
    if (length(overlap) > 0L) {
       stop("Attempt to overwrite generated parameters: ",
             paste(overlap, collapse=", "))
    }
    ## Filter extra so that any column where all numbers are with eps
    ## of the default strategy are not replaced:
    if (filter) {
      if (nrow(extra) == 0L) {
         extra <- NULL
       } else {
         pos <- diff(apply(extra, 2, range)) == 0
         if (any(pos)) {
            eps <- sqrt(.Machine$double.eps)
           x1 \leftarrow \text{extra}[1, \text{pos}]x2 \leftarrow \text{unlist}(s[\text{names}(x1)])drop <- abs(x1 - x2) < eps & abs(1 - x1/x2) < eps
           if (any(drop)) {
             keep <- setdiff(colnames(extra), names(drop)[drop])
              extra <- extra[, keep, drop=FALSE]
            }
         }
      }
    }
    if (!is.null(extra)) {
      m <- cbind(m, extra)
    }
    m
  }
<environment: 0x7f9512489620>
```
}

Table S1. Core parameter of the FF16 physiological model.

Table S2. Parameters for hyper-parameterisation of the FF16 physiological model.

Description	Symbol	Unit	Code	Value
Leaf turnover				
Global average leaf mass per area	ϕ_0	$\mathrm{kg}\,\mathrm{m}^{-2}$	lma_0	0.1978791
Rate of leaf turnover at average leaf mass per unit leaf area, ϕ_0	β_{k11}	\rm{vr}^{-1}	B_kl1	0.4565855
Scaling exponent for ϕ in leaf turnover	β_{k12}		B_k12	1.71
Sapwood turnover				
Global average wood density	ρ_0	$\mathrm{kg\,m^{-3}}$	rho_0	608
Rate of sapwood turnover at average wood density, ρ_0	$\beta_{\rm ks1}$	\rm{vr}^{-1}	B_ks1	0.2
Scaling exponent for ρ in sapwood turnover	$\beta_{\rm ks2}$		$B_k s2$	1.25
Photosynthesis				
Leaf nitrogen per unit leaf area	ν	$\mathrm{kg}\,\mathrm{m}^{-2}$	narea	0.00187
Global average nitrogen per unit leaf area	ν_0	$\mathrm{kg}\,\mathrm{m}^{-2}$	narea_0	0.00187
Potential CO ₂ photosynthesis at average leaf nitrogen, ν_0	$\beta_{\rm lf1}$	mold ⁻¹ m ⁻²	B_l	0.8273474
Curvature of light response curve	β_{lf2}		B_1f2	0.75
Quantum yield of leaf photosynthesis $(CO2)$ per unit photosynthetically active radi-	β_{lf3}		B_1f3	0.04
ation)				
Scaling exponent for leaf nitrogen in maximum leaf photosynthesis	$\beta_{\mathrm{lf}5}$		B_1f5	0.5
Respiration				
$CO2$ respiration per unit leaf nitrogen	$\beta_{\text{lf}4}$	mol yr ⁻¹ kg ⁻¹	B_1f4	15750
$CO2$ respiration per unit sapwood volume	β_{rs1}	mol yr $^{-1}$ m $^{-3}$	B_rs1	4012
$CO2$ respiration per unit bark volume	$\beta_{\rm rb1}$	mol yr ^{-1} m ^{-3}	B_rb1	8024

Fig. S1. Leaf turnover decreases with leaf-mass per unit leaf area. Data from [\(5\)](#page-12-5) for 678 species from 51 sites, each point giving a species-average. Lines show standardised major axis lines fitted to data from each site, with intensity of shading adjusted according to strength of the relationship.

Fig. S2. Assumptions of a functional balance model for plant construction. Each dot is a single plant from the BAAD [\(6\)](#page-12-6). Blue lines show standardised major axis lines fit to different species. The black line shows the relationship assumed in this paper, with slope given by the functional-balance assumptions in Table 4 from main text.

Fig. S3. Effect of four traits on stem-diameter growth rate for different-sized plants. Growth rates were simulated using the plant model, applying the trade-offs describing in Table 3. Each panel shows how growth is influenced by a different trait for plants of a given height, and across a series of canopy openness values from completely open (light blue, *E* = 1) to heavily shaded (dark line, *E* = 0*.*25). For any given value of trait and *E*, plants were grown to the desired stem-diameter and their growth rate estimated. Changes in trait-growth relationships are summarised in Table 5.

Fig. S4. Effect of four traits on stem-area growth rate for different-sized plants. Growth rates were simulated using the plant model, applying the trade-offs describing in Table 3. Each panel shows how growth is influenced by a different trait for plants of a given height, and across a series of canopy openness values from completely open (light blue, *E* = 1) to heavily shaded (dark line, *E* = 0*.*25). For any given value of trait and *E*, plants were grown to the desired stem-area and their growth rate estimated. Changes in trait-growth relationships are summarised in Table 5.

Fig. S5. Effect of four traits on plant mass growth rate for different-sized plants. Growth rates were simulated using the plant model, applying the trade-offs describing in Table 3. Each panel shows how growth is influenced by a different trait for plants of a given height, and across a series of canopy openness values from completely open (light blue, $E = 1$) to heavily shaded (dark line, $E = 0.25$). For any given value of trait and E , plants were grown to the desired mass and their growth rate estimated. Changes in trait-growth relationships are summarised in Table 5.

References

- 1. Falster DS, FitzJohn RG, Brännström Å, Dieckmann U, Westoby M (2016) plant: A package for modelling forest trait ecology and evolution. *Methods in Ecology and Evolution* 7:136–146.
- 2. R Core Team (2015) *R: a language and environment for statistical computing* (R Foundation for Statistical Computing, Vienna, Austria).
- 3. Falster DS, Brännström Å, Dieckmann U, Westoby M (2011) Influence of four major plant traits on average height, leaf-area cover, net primary productivity, and biomass density in single-species forests: a theoretical investigation. *Journal of Ecology* 99(1):148–164.
- 4. Henery M, Westoby M (2001) Seed mass and seed nutrient content as predictors of seed output variation between species. *Oikos* 92:479–490.
- 5. Wright IJ, et al. (2004) The world-wide leaf economics spectrum. *Nature* 428(2403):821–827.
- 6. Falster DS, et al. (2015) BAAD: a Biomass And Allometry Database for woody plants. *Ecology* 96(5):1445.