

The worldwide leaf economics spectrum

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Bringing together leaf trait data spanning 2,548 species and 175 sites we describe, for the first time at global scale, a universal spectrum of leaf economics consisting of key chemical, structural and physiological properties. The spectrum runs from quick to slow return on investments of nutrients and dry mass in leaves, and operates largely independently of growth form, plant functional type or biome. Categories along the spectrum would, in general, describe leaf economic variation at the global scale better than plant functional types, because functional types overlap substantially in their leaf traits. Overall, modulation of leaf traits and trait relationships by climate is surprisingly modest, although some striking and significant patterns can be seen. Reliable quantification of the leaf economics spectrum and its interaction with climate will prove valuable for modelling nutrient fluxes and vegetation boundaries under changing land-use and climate.

Green leaves are fundamental for the functioning of terrestrial ecosystems. Their pigments are the predominant signal seen from space. Nitrogen uptake and carbon assimilation by plants and the decomposability of leaves drive biogeochemical cycles. Animals, fungi and other heterotrophs in ecosystems are fuelled by photosynthate, and their habitats are structured by the stems on which leaves are deployed. Plants invest photosynthate and mineral nutrients in the construction of leaves, which in turn return a revenue stream of photosynthate over their lifetimes. The photosynthate is used to acquire mineral nutrients, to support metabolism and to re-invest in leaves, their supporting stems and other plant parts.

There are more than 250,000 vascular plant species, all engaging in the same processes of investment and reinvestment of carbon and mineral nutrients, and all making enough surplus to ensure continuity to future generations. These processes of investment and re-investment are inherently economic in nature^{1–3}. Understanding how these processes vary between species, plant functional types

and the vegetation of different biomes is a major goal for plant ecology and crucial for modelling how nutrient fluxes and vegetation boundaries will shift with land-use and climate change.

Data set and parameters

We formed a global plant trait network (Glopnet) to quantify leaf economics across the world's plant species. The Glopnet data set spans 2,548 species from 219 families at 175 sites (approximately 1% of the extant vascular plant species). The coverage of traits, species and sites is at least tenfold greater than previous data compilations^{4–11}, extends to all vegetated continents, and represents a wide range of vegetation types, from arctic tundra to tropical rainforest, from hot to cold deserts, from boreal forest to grasslands. Site elevation ranges from below sea level (Death Valley, USA) to 4,800 m. Mean annual temperature (MAT) ranges from -16.5°C to 27.5°C ; mean annual rainfall (MAR) ranges from 133 to 5,300 mm per year. This covers most of the range of MAT–MAR space in which higher plants occur¹² (Fig. 1). The broad coverage of the data set has

allowed us to quantify the relationships of leaf economics to climate at a scale not previously possible. Here we report some global outcomes from our analyses.

We focus on six key features of leaves that together capture many essentials of leaf economics. (1) Leaf mass per area (LMA) measures the leaf dry-mass investment per unit of light-intercepting leaf area deployed. Species with high LMA have a thicker leaf blade or denser tissue, or both. (2) Photosynthetic assimilation rates measured under high light, ample soil moisture and ambient CO₂ are here called photosynthetic capacity (A_{mass}) for brevity. Photosynthetic capacity is influenced both by stomatal conductance and by the drawdown of CO₂ concentration inside the leaf (carboxylation capacity). (3) Leaf nitrogen (N) is integral to the proteins of photosynthetic machinery, especially Rubisco^{8,13}. The photosynthetic machinery is responsible for drawdown of CO₂ inside the leaf, a process also affected by leaf structure^{14,15}. (4) Leaf phosphorus (P) is found in nucleic acids, lipid membranes and bioenergetic molecules such as ATP. Phosphorus derives from weathering of soil minerals at a site, in contrast to nitrogen, much of which may be fixed from the atmosphere by plants. (5) Dark respiration rate (R_{mass}) reflects metabolic expenditure of photosynthate in the leaf, especially protein turnover and phloem-loading of photosynthates¹⁶. (6) Leaf lifespan (LL) describes the average duration of the revenue stream from each leaf constructed. Long LL requires

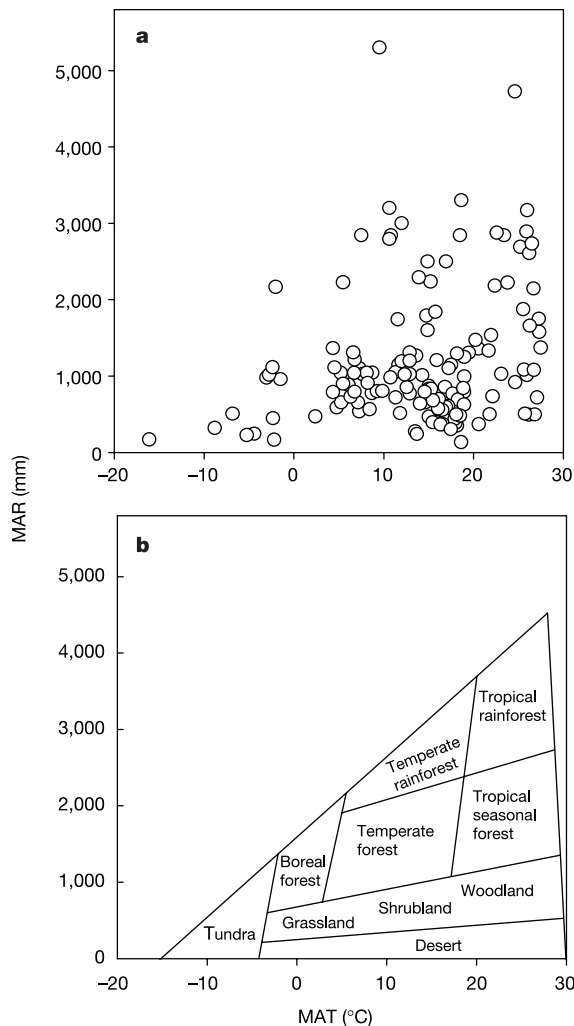


Figure 1 Mean annual rainfall (MAR) and mean annual temperature (MAT). Results for the 175 sites from where leaf data were compiled (a), in relation to major biome types of the world (b), following ref. 12. Biome boundaries are only approximate.

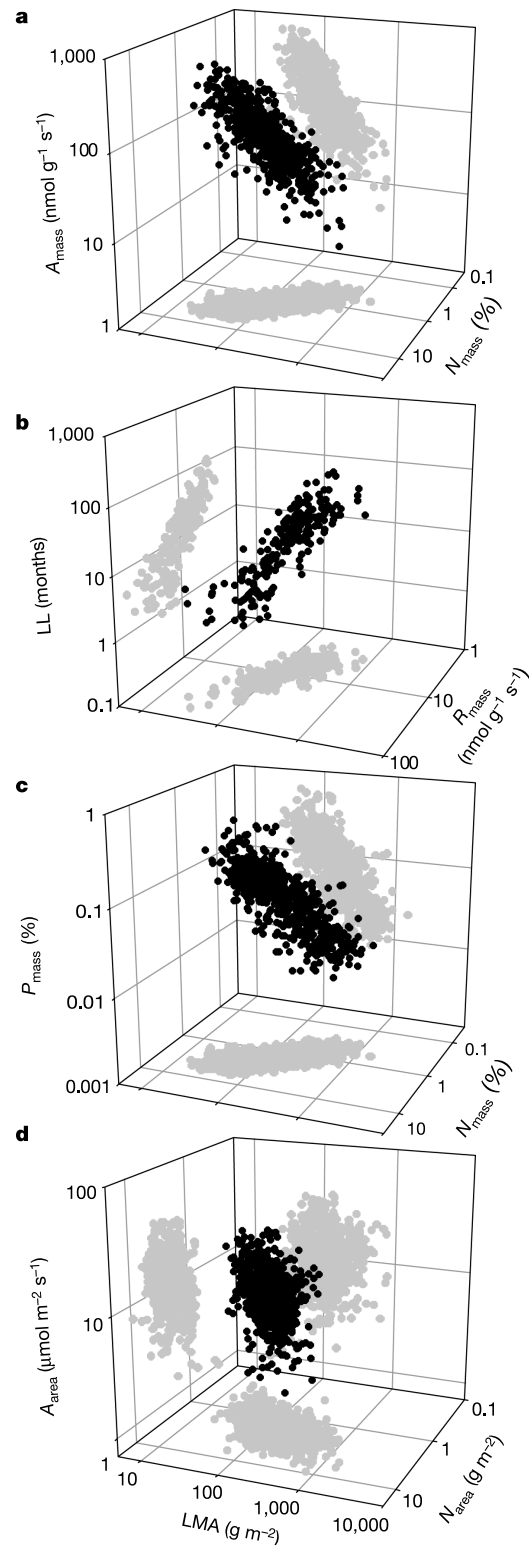


Figure 2 Three-way trait relationships among the six leaf traits with reference to LMA, one of the key traits in the leaf economics spectrum. The direction of the data cloud in three-dimensional space can be ascertained from the shadows projected on the floor and walls of the three-dimensional space. Sample sizes for three-way relationships are necessarily a subset of those for each of the bivariate relationships. **a**, A_{mass} , LMA and N_{mass} ; 706 species. **b**, LL, R_{mass} and LMA; 217 species. **c**, N_{mass} , P_{mass} and LMA; 733 species. **d**, A_{area} , LMA and N_{area} ; 706 species.

robust construction in the form of high LMA^{6,17,18}.

Ecophysiological attributes such as leaf N, leaf P, A_{mass} or R_{mass} can be expressed per leaf area, per leaf dry mass or per leaf volume. A leaf-area basis reflects fluxes in relation to surfaces. It is a natural basis for describing light capture and for expressing transactions through surfaces such as trade-offs between carbon gain and water transpiration^{19,20}. On a mass basis, leaf economics are quantified in terms of revenues and expenditures per unit investment, measured as biomass, or C, N or P. Scaling up to whole plants, mass-based expressions of leaf nutrient concentrations (N_{mass} , P_{mass}), A_{mass} and R_{mass} are more tightly correlated than area-based expressions to relative growth rates of seedlings or to absolute height growth rates of young trees^{21,22}. By definition, area- and mass-based traits can be interconverted via LMA (for example, $N_{\text{area}} = N_{\text{mass}} \times \text{LMA}$). Here we report relationships among leaf traits using both area- and mass-based formulations, and also analyses where the effect of LMA is explicitly controlled.

We first quantify relationships among the six mass-based leaf traits: A_{mass} , R_{mass} , LMA, LL, N_{mass} and P_{mass} . We find that these economic traits covary tightly across all species. Trait relationships are similar for species pooled by growth form, plant functional group or biome, which indicates the existence of a single global spectrum of leaf economic variation. Second, we treat photosynthetic capacity, R_{mass} and leaf nutrient concentrations on a leaf-area basis. These relationships are not as strong as among mass-basis traits, and we consider why. Third, we assess the influence of climate on leaf trait relationships. We find that, in general, the influence is modest, although particular traits and trait relationships show substantial patterning with climate.

The leaf economics spectrum

Mass-based leaf traits

The six mass-basis leaf traits varied by one to two orders of magnitude across the data set. LMA ranged from 14 to 1,500 g m⁻² and LL from 0.9 up to 288 months. A_{mass} ranged from 5 to 660 nmol g⁻¹ s⁻¹; dark respiration from 2.2 to 65 nmol g⁻¹ s⁻¹. N_{mass} ranged from 0.2 to 6.4%; P_{mass} from 0.008 to 0.6%. Considered pairwise, all leaf traits were highly correlated (Table 1). These correlations have been reported previously from smaller data sets^{6-10,23}. Here we have generalized the patterns over many more species, sites and vegetation types.

We moved beyond pairwise consideration of traits to determine the extent to which leaf economic traits covary in multidimensional trait space. This covariation can be quantified as the proportion of total trait variation explained by the first principal axis in a principal components analysis. In two-trait space, the principal axis is the long axis of the ellipse resulting from two correlated traits. In three-trait space, the principal axis is the long axis of an ellipsoid. In multidimensional trait space, the principal axis describes the main axis of variation through a hyperellipsoid²⁴. A remarkable 82% of all variation in A_{mass} , LMA and N_{mass} across species lay along the first principal axis in three-trait space (Fig. 2a). Because some of the residual 18% must be measurement variation, 82% represents a minimum estimate of the dominance of this single spectrum in explaining variation across plant species worldwide. Further three-

dimensional subsets of the six-dimensional data set are shown in Figs 2b and c. Multi-dimensional analyses including from four to all six of the traits similarly showed the large majority of variation explained with a single axis (Table 2). With the six traits included, 74% of all variation lay along the first principal axis.

The extent to which each trait contributed to the principal axis of variation is indicated by a loading (or weight) assigned to each trait. The directionality of these loadings (Table 2) indicates that the axis can be thought of as a leaf economics spectrum. This spectrum runs from species with potential for quick returns on investments of nutrients and dry mass in leaves to species with a slower potential rate of return. At the quick-return end are species with high leaf nutrient concentrations, high rates of photosynthesis and respiration, short leaf lifetimes and low dry-mass investment per leaf area. At the slow-return end are species with long leaf lifetimes, expensive high-LMA leaf construction, low nutrient concentrations, and low rates of photosynthesis and respiration.

Within growth forms or functional groups the principal axes of variation had the same directionality of trait correlations as for the total data set (Table 2). Similarly, species grouped by major biome type (Fig. 1), or by MAT or MAR classes, yielded the same pattern (data not shown). The concordance of these results is of special significance, indicating a coordination of these key leaf traits that is consistent across major plant functional types, growth forms and biomes. The amount of variation captured by the principal axis in the different species groupings was also similar to that across all species in most cases.

The main exception was among deciduous trees and shrubs where, as expected, there was substantially less variation in LL (5-fold versus 100-fold), and where LL–LMA relationships were partially uncoupled. Still, whereas different growth forms and functional groups were differentiated along the leaf economics spectrum when trait means were considered, the overlap between species groups was large (data not shown). Evergreen trees and shrubs had longer mean LL and higher LMA than deciduous species, but evergreens had much wider ranges for both traits, extending to LLs almost as short as for the shortest-LL deciduous species, and to similarly low LMA. Similarly, on average, shrubs and trees had higher LMA and longer LLs but lower N_{mass} , A_{mass} and R_{mass} than herbs and grasses, yet trees and shrubs spanned almost the entire range of any of these leaf traits. Another example: N₂-fixing species had higher mean N_{mass} than non N₂-fixing plants, yet the range of N_{mass} was larger and extended higher in non N₂-fixing species.

Allometries among traits

Slopes on log–log axes, or ‘scaling exponents’, indicate the proportionality of pairwise trait relationships. Most slopes were significantly different from +1.0 or –1.0 (Table 1, above diagonal). In other words, the traits showed allometric relationships rather than scaling in direct proportion with one another (‘isometry’). A tenfold increase in P_{mass} corresponded with a 4.7-fold increase in N_{mass} (scaling slope 0.67), indicating that N:P ratios decline as one moves towards the end of the spectrum that represents quick returns on investments of carbon and nutrients. The stoichiometry between

Table 1 Mass basis of bivariate relationships between the leaf traits

	log LL	log LMA	log A_{mass}	log N_{mass}	log P_{mass}	log R_{mass}
log LL		1.71 (1.62, 1.82)	-1.38 (-1.45, -1.31)	-2.26 (-2.39, -2.14)	-1.06 (-1.19, -0.94)	-1.67 (-1.82, -1.53)
log LMA	0.42 (678)		-0.75 (-0.79, -0.72)	-1.28 (-1.32, -1.24)	-0.82 (-0.86, -0.78)	-0.96 (-1.05, -0.88)
log A_{mass}	0.68 (512)	0.50 (764)		1.72 (1.63, 1.81)	1.03 (0.91, 1.16)	1.18 (1.09, 1.27)
log N_{mass}	0.42 (706)	0.57 (1958)	0.53 (712)		0.66 (0.64, 0.69)	0.70 (0.64, 0.76)
log P_{mass}	0.24 (207)	0.55 (739)	0.16 (212)	0.72 (745)		1.04 (0.87, 1.25)
log R_{mass}	0.60 (217)	0.45 (274)	0.59 (259)	0.55 (267)	0.34 (78)	

Standardized major axis slopes with 95% confidence intervals are given in the upper right section of the matrix (y variable is column 1, x variable in row 1). Coefficients of determination (r^2) and sample sizes are given in the lower left section of the matrix. All relationships were highly significant, $P \ll 0.0001$. Further details allowing calculation of predictive regression equations for each pair of leaf traits are given in Supplementary Information.

Table 2 Principal components analyses of global leaf trait data

	All species	Trees*	Shrubs	Herbs*	Grasses*	N ₂ -fixers*	Non N ₂ -fixers	C3	C4†	Broad-leaved shrubs and trees	Needle-leaved shrubs and trees*	Evergreen shrubs and trees	Deciduous shrubs and trees*
Variation explained (%)	74.4	77.2	72.0	67.8	70.7	79.6	75.0	73.0	68.7	72.4	58.0	64.0	48.5
Leaf trait	Loadings												
LL	-0.85	-0.90	-0.83	-0.73	-0.86	-0.86	-0.84	-0.88		-0.84	-0.79	-0.74	-0.43
LMA	-0.88	-0.84	-0.84	-0.72	-0.91	-0.87	-0.87	-0.84	-0.84	-0.84	-0.73	-0.82	-0.28
N _{mass}	0.91	0.82	0.91	0.89	0.71	0.88	0.92	0.90	0.71	0.88	0.71	0.87	0.81
A _{mass}	0.86	0.91	0.82	0.88	0.79	0.96	0.86	0.88	0.87	0.83	0.82	0.76	0.84
P _{mass}	0.88	0.93	0.83	0.89	0.93	0.89	0.89	0.90	0.88	0.84	0.74	0.82	0.90
P _{mass}	0.79		0.87				0.80	0.72		0.85		0.78	

The principal axis or component explained 74.4% of variation in the total data set. LL and LMA were negatively correlated with this primary axis of variation while the other traits were positively correlated with it, both in the total data set and in data subsets defined by growth form or functional group. The same directionality of trait loadings and similarly high percentage of variance was explained by the principal axis with species grouped by site temperature, rainfall or altitude, or with sites grouped into major biome type following Fig. 1 (data not shown), demonstrating the broad generality of the coordinated spectrum of leaf economics. All data were log₁₀-transformed before analyses.

*P_{mass} excluded due to too few data.

†P_{mass} and LL excluded due to few data. C3, C4 indicate species with C3 and C4 photosynthetic pathways, respectively.

leaf N and P has begun to attract increasing interest as an index of soil nutrient limitation (taken across sets of co-occurring species)²⁵, and also because it relates to plant growth strategies and influences plant–herbivore interactions in food webs²⁶.

The slope of the LL–LMA relationship was 1.7, meaning that tenfold greater dry mass invested per unit leaf area coincided with 50-fold longer LL. All else being equal, this implies that the measure of light-intercepting leaf area (in mm²) × duration (in months) per gram leaf was greater for high-LMA than for low-LMA species. If this measure translated directly into fitness benefit, this might lead to runaway selection towards ever-increasing LMA and LL²⁷. But continuing ecological success of low-LMA species shows that all else is not equal. On average, a tenfold decrease in LMA, for example, coincided with a 21-fold increase in photosynthetic capacity. Further, low LMA, high A_{mass} and generally faster turnover of plant parts permit a more flexible response to the spatial patchiness of light and soil resources²⁸, as well as conferring advantages via a compound interest effect, whereby carbon fixed earlier can be reinvested in new leaves sooner^{27,29}. On the other hand, high A_{mass} requires high N_{mass}, and the combination of high LMA and high N_{mass} may increase vulnerability to herbivory as well as increasing energy losses via respiration⁹, which can be detrimental in situations where energy gain is low owing to low resource availability, such as under low light conditions³⁰.

Area versus mass basis of expression

Because N_{mass} = N_{area}/LMA, we considered whether the relationship between N_{mass} and LMA should be thought of as arising from N_{area} values that do not vary greatly across species, divided through by highly varying LMA. The evidence contradicts this interpretation. First, N_{area} varied more widely than did N_{mass} (35-fold versus 26-fold; data for 1,958 species). Second, the -1.28 slope of the N_{mass}–LMA relationship was significantly steeper (Table 1) than the -1.0 slope expected if N_{area} was independent of LMA.

N_{area} was indeed correlated with LMA, but positively, and more weakly than were N_{mass} and LMA (r² = 0.34 versus 0.57; Table 3). Gradients of leaf N on a mass versus an area basis hence represent

fundamentally different multiple-trait gradients because of their different patterns of covariation with LMA³¹. The leaf economics spectrum of species from low to high N_{mass} also constitutes a spectrum of decreasing LMA and LL, and of increasing A_{mass}, R_{mass} and P_{mass}. But a spectrum of leaf types in terms of N_{area} would be less informative. A given N_{area} can result from low N_{mass} combined with high LMA, high N_{mass} combined with low LMA, or from combinations in between. In general, low N_{mass} with high LMA represents a species with long-lived leaves and low A_{mass}, while high N_{mass} with low LMA represents the opposite. As a result, a wider variety of leaf types may be found at a given N_{area} than at a given N_{mass}^{31–33}. Here for example, LMA varied approximately 20-fold at the grand mean of N_{area} versus tenfold at the grand mean of N_{mass}.

Because of the covariation between leaf N and LMA, relationships between leaf N and other traits changed substantially when expressed on an area rather than on a mass basis. As seen previously^{7–9,31,33}, relationships between leaf N and A_{mass} or dark respiration were weaker when considered on an area basis, as were relationships between leaf P and other traits (Table 3). By also including LMA in analyses, we can quantify the independent effects of leaf structure and nutrient content on A_{mass} and dark respiration^{32,34}. A_{mass} increased with increasing N_{mass} at any given LMA, and decreased with increasing LMA at any given N_{mass} (partial regression coefficients for LMA and leaf N, P < 0.001; regression details in Supplementary Information). Similarly, A_{area} increased with increasing N_{area} at any given LMA, and decreased with increasing LMA at any given N_{area} (Fig. 2d and Supplementary Information). That is, both leaf structure and nitrogen concentration affect photosynthetic capacity^{5,6,8,32}. The independent LMA effect is most probably due to leaves with high mass per area having longer diffusion paths from stomata to chloroplasts or greater internal shading of lower chloroplasts, limiting the A_{mass} possible for a given leaf protein content^{14,15,35}. Also, less of the N may be invested in photosynthetic versus non-photosynthetic leaf components in high-LMA species^{36,37}. Similarly, N_{mass} and LMA showed independent effects on R_{mass} (partial regression coefficients both

Table 3 Area basis of bivariate relationships between the six leaf traits

	log LL	log LMA	log A _{area}	log N _{area}	log P _{area}	log R _{area}
log LL		1.71 (1.62, 1.82)	-2.12 (-2.30, -1.96)	-2.36 (-2.19, -2.54)	NA (P = 0.862)	NA (P = 0.211)
log LMA	0.42 (678)		NA (P = 0.152)	1.54 (1.48, 1.59)	1.23 (1.14, 1.32)	1.20 (1.08, 1.34)
log A _{area}	0.13 (512)	0.003 (764)		1.21 (1.13, 1.31)	0.66 (0.58, 0.75)	0.93 (0.83, 1.04)
log N _{area}	0.04 (706)	0.34 (1,958)	0.13 (722)		0.68 (0.64, 0.72)	0.89 (0.80, 0.98)
log P _{area}	0.0002 (202)	0.01 (739)	0.02 (223)	0.35 (750)		1.26 (1.04, 1.54)
log R _{area}	0.01 (217)	0.14 (274)	0.19 (259)	0.34 (267)	0.26 (78)	

Standardized major axis slopes with 95% confidence intervals are given in the upper right section of the matrix (y variable is column 1, x variable in row 1). Coefficients of determination (r²) and sample sizes are given in the lower left section of the matrix. NA, not applicable: in these cases the correlation was clearly non-significant (P values given in parentheses). While a standardized major axis can still be fitted in such cases, its slope is essentially meaningless. All other relationships were highly significant, P < 0.0001, with the exception of those between log P_{area} and each of log LMA (P = 0.029) and log A_{area} (P = 0.034). Further details allowing calculation of predictive regression equations for each pair of traits are given in the Supplementary Information.

$P < 0.001$), although the physiological basis for an independent LMA effect on R_{mass} is less clear than for photosynthetic capacity⁹.

Relating to the weaker and somewhat different pattern of area-based trait relationships, the principal axis of a principal components analysis involving LMA, LL, A_{area} , R_{area} , N_{area} and P_{area} explained a lower proportion of total trait variation than the mass-based equivalent and had a different pattern of trait loadings. This principal axis explained 43% of variation (versus 74% on a mass basis), and largely reflected a spectrum of increasing N_{area} (the trait with the strongest trait loading; Supplementary Information). The five other traits showed positive loadings with the principal axis, with these loadings clearly weaker than in the mass-based analysis. About half the residual variation was explained by a second principal axis, which expressed essentially the same trait correlations as the first axis from the mass-based analysis. Together, these first two axes explained 72% of total trait variation in the area-based data set, marginally less than the first principal axis in the mass-based analysis. Area-based analyses for species grouped by growth form, functional type or biome showed patterns broadly concordant with those across all species, though less clearly than for the mass-based analyses (results not shown).

In summary, the coordination among leaf traits appears to be stronger and simpler on a mass basis than an area basis. This is not because area-basis traits are less varied among species. Rather it is because the LMA–LL spectrum is related to mass-based nutrient concentrations and assimilation and respiration rates in a simpler way than on an area basis.

Climate influence on leaf investment

Plant ecologists have emphasized broad relationships between leaf traits and climate for at least a century. In particular, a general tendency for species inhabiting arid and semi-arid regions to have leathery, high-LMA leaves has been reported^{4,10,38–40}. Building high-LMA leaves needs more investment per unit leaf area. Construction cost per unit leaf mass varies relatively little between species: leaves with high protein content (typically low-LMA leaves) tend to have low concentrations of other expensive compounds such as lipids or lignin, and high concentrations of cheap constituents such as minerals⁴¹. Leaf traits associated with high LMA (for example, thick leaf blade; small, thick-walled cells) have been interpreted as adaptations that allow continued leaf function (or at least postpone leaf death) under very dry conditions, at least in evergreen species.

We characterized sites by annual means of temperature, potential evapotranspiration (PET), vapour pressure deficit (VPD) and solar

irradiance, and MAR. PET, VPD and irradiance were cross-correlated with rainfall and temperature. Against expectations, LMA showed only a very weak relationship with lower rainfall, considered worldwide ($r^2 = 0.002$, $P = 0.032$; 2,370 species from 163 sites). LMA was actually more strongly (positively) correlated with MAT ($r^2 = 0.10$). Worldwide, precipitation is correlated with temperature, cold high-latitude environments typically having low precipitation (Fig. 1). Once variation in MAT was controlled in a multiple regression, LMA did indeed increase as rainfall decreased (Fig. 3). Similarly, LMA was more strongly (positively) correlated with VPD ($r^2 = 0.15$), PET ($r^2 = 0.15$) or site irradiance ($r^2 = 0.18$) than with MAT or site rainfall alone.

Despite the substantial overlap in leaf traits of evergreen and deciduous species, these species groups varied somewhat in their leaf trait–climate relationships. LMA and rainfall showed a strong negative relationship in evergreen shrubs and trees, whereas in deciduous species they were virtually unrelated ($r^2 = 0.22$ versus 0.002). Across all species, LL was positively correlated with PET, VPD, MAT and site irradiance (r^2 values ranging from 0.04 to 0.10). In deciduous trees and shrubs these relationships were consistently stronger (r^2 values ranged from 0.37 to 0.51). That is, LL of deciduous species was shorter at colder sites where the growing season was shorter. But in evergreen shrubs and trees, LL tended to be longer at colder, lower-humidity sites (for example, MAT versus LL, $r^2 = 0.10$; PET versus LL, $r^2 = 0.18$). This is consistent with cold climate vegetation being typically N-limited and demonstrating ‘slow-return’ strategies⁴².

Trait coordination is largely independent of climate

A major aim of the Glopnet collaboration was to obtain enough coverage of climate variation to dissect out effects of climate on relationships between leaf economic traits. There were indeed statistically significant effects of climate. Nevertheless, a major finding from this project is that the influence of climate was, in general, quite modest. How can this be, given that traits such as LMA and LL vary systematically with MAR, MAT and other climate indices? The answer seems to have two elements.

First, much of the total leaf economic variation occurs among co-existing species. The proportion of total variation in LL within sites was 57%, the remaining 43% occurring between sites (variance components analysis). For R_{mass} , the proportion of within-site variation was 67%, for A_{mass} 48%, for N_{mass} 38%, for LMA 36%,

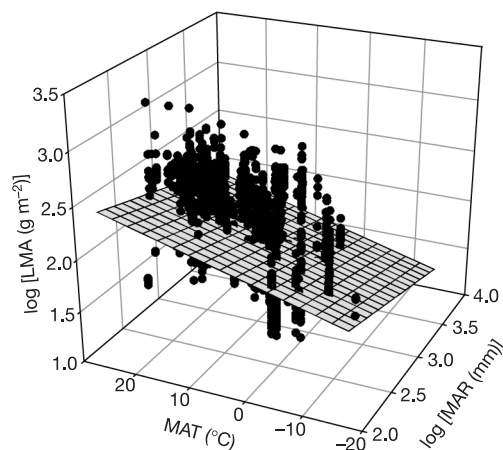


Figure 3 LMA as a function of MAT and MAR at the study sites (data for 2,370 species from 163 sites; rainfall and LMA are \log_{10} -scaled). The coefficients for MAT and log rainfall were highly significant in a multiple regression (both $P < 0.0001$; further details given in Supplementary Information).

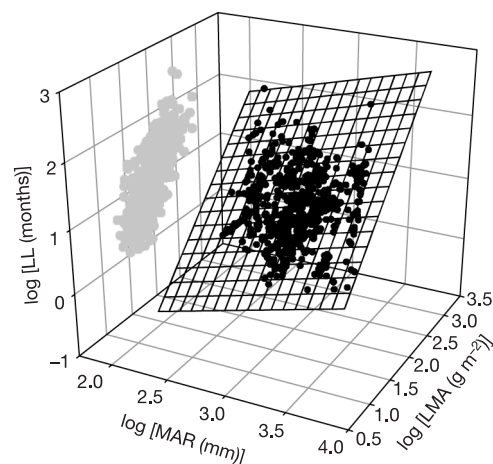


Figure 4 LL as a function of LMA and MAR (all axes are \log_{10} -scaled). When viewed in three dimensions, the two-dimensional LL–LMA cloud of points is spread along a sloping surface. The slope of this surface is steeper in the LMA dimension than in the rainfall dimension, reflecting the higher partial regression coefficient for LMA (1.23 versus 0.47). Both coefficients were highly significant in a multiple regression ($P < 0.0001$; $r^2 = 0.51$; data for 678 species from 51 sites).

for P_{mass} 20%. Similar or higher proportions of within-site variation were seen for area-based traits (40–57% of total variation). Second, leaf traits tend to vary in concert, with the leaf economics spectrum operating similarly in different biomes. Considered across all species, when single or multiple climate variables were included in regressions of LMA on N_{mass} , of A_{mass} on N_{mass} , or of LL on A_{mass} , they added a maximum of 0.05 to the r^2 . Similar results were found for most other bivariate trait relationships, as well as for regression models predicting one leaf trait from two or more other leaf traits as well as climate variables (<15% explanatory power added).

This is not to say that climate does not exert important influences on trait relationships. With such large sample sizes, these effects were still highly statistically significant. Rather, it is an extension of the fact that so much of the total variation in leaf traits is captured by the primary axis of the leaf economics spectrum. Adding climate variables to regressions involving area-based traits also added little explanatory power, for the most part. Exceptions were cases where the bivariate trait relationship was particularly weak to start with (for example, between A_{area} and LMA, R_{area} and LMA, or LL and A_{area}), climate variables added 0.15 to 0.28 to the model r^2 .

LL–LMA relationships shift with climate

The relationship between LMA and LL is important for leaf economics, LMA reflecting the dry-mass cost of deploying new leaf area, and LL representing the duration over which photosynthetic revenue is returned. LL increased with LMA across all species, LMA explaining 42% of variation in LL (Table 1). This relationship showed quite substantial patterning with climate. At a given LMA, LL was shorter at sites of lower rainfall (Fig. 4), meaning that the duration of return from a unit investment in leaf tissue tends to be shorter in such places. In addition, at sites with harsher climate the gain in LL for a given increase in LMA was less. Significant negative interaction terms in the regression models (all $P < 0.001$) indicated decreasing steepness for LL–LMA relationships in hotter, less humid (higher VPD or PET) or higher irradiance environments. The patterning of LL–LMA relationships with rainfall that was seen in our global data set was reported previously for evergreen species from four sites in eastern Australia¹⁷.

Gloynet results serve to confirm and extend the pattern across several hundred more species, from many more vegetation types. The previous comparisons deliberately contrasted sites with similar MAT. Here, by taking data from many more sites, representing most of the major biomes of the earth, we have been able to identify previously undescribed patterning of LL–LMA relationships with respect to climate, that is, the shallower response of LL to LMA with increasing MAT, VPD, PET and site irradiance.

A possible reason why a given LMA is associated with shorter LL in drier environments involves leaf nitrogen. For evergreen species in semi-arid habitats in eastern Australia¹⁰, the CO_2 concentration inside leaves during photosynthesis was lower at a given stomatal conductance to CO_2 (g_{CO_2}) than for evergreen species at higher rainfall sites. That is, they had higher carbon fixation (A_{area}) at a given stomatal conductance to water per unit leaf area ($g_{\text{H}_2\text{O}}$), because A_{area} is largely the product of CO_2 drawdown and g_{CO_2} , and g_{CO_2} is directly proportional to $g_{\text{H}_2\text{O}}$ (ref. 13). The lower CO_2 concentration inside leaves in the species from semi-arid sites was linked with higher N_{area} , but the high N_{area} corresponded to less robust tissue—a larger proportion of photosynthetic mesophyll—resulting in leaves that were less strong in a biomechanical sense at a given LMA and, it was argued, leading to the observed shorter LL at a given LMA¹⁷.

We do not yet have leaf toughness information or leaf anatomy widely across our global data set, but N_{area} did indeed increase with decreasing site rainfall in evergreen trees and shrubs ($r^2 = 0.18$), as well as across all species (albeit more weakly; $r^2 = 0.04$), CO_2 drawdown increased with increasing N_{area} at a given $g_{\text{H}_2\text{O}}$ (see

Supplementary Information), and A_{area} was higher at a given $g_{\text{H}_2\text{O}}$ at lower rainfall sites (see Supplementary Information), indicating that CO_2 concentration inside leaves tended to be lower at low rainfall sites (correlation $r = 0.44$, $P < 0.0001$). Taken together, these trends can be understood as resulting from the simultaneous optimization of nitrogen- and water-use during photosynthesis^{19,20}. They suggest that similar mechanisms operate in many of the world’s vegetation types, as were seen in the eastern Australian species.

Conclusions

We now have wide-ranging and convincing evidence that feasible leaf investment strategies are to a great extent arrayed along a single spectrum, with the same patterning of trait correlations seen globally and in species grouped by growth form, biome or climate. Besides leaf economics, many activities such as seed production, root economics and relations with mycorrhizas are undoubtedly important for plant fitness. But, surely, the broad generality of the relationships we describe suggests that natural selection eventually eradicates leaf investment strategies that are not economically competitive. The leaf economics spectrum reflects a mixture of direct and indirect causal relationships between traits.

For example, the linkage of high A_{mass} with high N_{mass} is in large part the result of a direct causal relationship⁸. Similarly, long LL requires the robustness and low palatability (including chemical defences) associated with high LMA^{17,18}. More indirectly, high A_{mass} tends to be associated with short LL because it requires high N_{mass} and/or low LMA, which increase leaf vulnerability to herbivory and physical hazards, and because high A_{mass} drives fast growth, rapidly shading older leaves, leading them to senesce once their resources become more valuable when transferred to better-lit newer foliage⁴³. The absence of outliers in the trait relationships is particularly striking. Although species vary widely in growth form, life history and niche space occupied, it seems that a mixture of physiological causation and the demands of competitiveness constrain species data points within tightly bounded domains of trait space⁶.

Overall, we found that the effect of climate on leaf trait relationships was modest, although particular trait-pairs showed striking and significant patterns with site climatic properties. Reliable quantification of the global leaf economics spectrum and its relationship to climate should prove valuable in modelling how carbon and nitrogen are tied up and used in plant biomass, and how nutrient fluxes and vegetation boundaries will shift with land-use and climate change. Recent models of vegetation dynamics under global change^{44,45} do make use of the LMA spectrum previously described in ref. 6 for the purpose of grouping species into plant functional types. Recognition of the leaf economics spectrum strengthens and consolidates our knowledge in this regard. At the same time, it is important to note that the spectrum is continuous, rather than divided into distinct categories either by growth form or by habitat. Formulations of plant functional typologies that represent variation as a continuous spectrum, rather than distinct categories, bear promise for the future. □

Methods

Leaf traits

Leaf trait data were compiled from both published and unpublished sources. A data set was considered suitable provided it contained data for at least two of the leaf traits for at least four co-occurring species. Highly artificial vegetation types such as forestry plantations and crop fields were not included. Only site-based data sets were used so that we could reasonably attach climate data. The total data set (Supplementary Information) represented 175 sites and contained 2,548 species–site combinations, consisting of 2,021 species, with 350 occurring at more than one site.

Climate

Climate data were taken from (1) the sites themselves, where known; (2) the nearest weather stations, with temperature data scaled where necessary by an altitudinal lapse rate of 0.6 °C per 100 m (ref. 46); (3) a global 0.5 × 0.5° data set of MAT, MAR, vapour pressure and irradiance⁴⁷; (4) a global 0.5 × 0.5° data set of Penman–Monteith PET calculated for

the period 1987–1988 (ref. 48). MAT and rainfall data from the global data set agreed closely with known or nearby-station data (scaled for altitude where necessary), giving us confidence that the vapour pressure and irradiance data were reliable also. Irradiance data were not adjusted for altitude because increases in cloud cover with elevation tend to offset the increase in radiation that would be observed in clear air, although this may not be true for high-elevation sites in arid regions^{46,49}. For high-elevation sites, PET was estimated from a regression equation fitted to all other sites, which considered PET as a function of MAT and annual rainfall ($r^2 = 0.71$). Vapour pressure was scaled for high-elevation sites using an empirical formula expressing the exponential decrease of vapour pressure with altitude⁴⁹. Monthly mean VPD was estimated as the difference between the saturation vapour pressure of air (at the monthly mean temperature) and vapour pressure taken from the global data set. Saturation vapour pressure was calculated using the Tetens formula⁵⁰. Climate variables were averaged or summed (rainfall) across all months of the year, and for those months with mean temperature over 4.99 °C, giving an estimate of climate during the growth season. Results using yearly and growth-season climate indices differed little, so we report results relating to yearly climate averages only.

Data transformation and analysis

For each species at each site, the mean value for each trait was used. Where traits were reported separately for sun leaves and shade leaves, sun-leaf data were used. Similarly, if data were presented separately for recently matured and old leaves, that of the recently matured leaves were chosen. Leaf traits were approximately log-normally distributed across the data set, as were rainfall and VPD. Accordingly, these variables were log₁₀-transformed before analysis. MAT, PET and solar radiation were left untransformed because their distribution across sites was approximately normal.

Standardized major axis slopes⁵⁴ with 95% confidence intervals were fitted to bivariate trait relationships because our aim was to describe the best-fit lines or central axes of these ‘scaling’ relationships. Ordinary multiple regression was used for analyses exploring the additional predictive power of climate variables on leaf trait relationships. Regressions were first run including interaction terms between climate variables and leaf traits; where the interaction was non-significant ($P < 0.05$), we re-ran models with main effects only. In several cases, principal components analyses run on data subsets defined by growth form or plant functional type had to be re-run with one or more leaf traits removed, owing to insufficient data. These cases are indicated in Table 2. Variance component analyses were based on the decomposition of analysis of variance (ANOVA) type I sums of squares. Principal components analyses, variance components and regression analyses were run in SPSS for Windows version 11.01.

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