

# Leaf venation network architecture coordinates functional trade-offs across vein spatial scales: evidence for multiple alternative designs

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#### Summary

 Variation in leaf venation network architecture may reflect trade-offs among multiple functions including efficiency, resilience, support, cost, and resistance to drought and herbivory. However, our knowledge about architecture-function trade-offs is mostly based on studies examining a small number of functional axes, so we still lack a more integrative picture of multidimensional trade-offs.

 Here, we measured architecture and functional traits on 122 ferns and angiosperms species to describe how trade-offs vary across phylogenetic groups and vein spatial scales (small, medium, and large vein width) and determine whether architecture traits at each scale have independent or integrated effects on each function.

 We found that generalized architecture-function trade-offs are weak. Architecture strongly predicts leaf support and damage resistance axes but weakly predicts efficiency and resilience axes. Architecture traits at different spatial scales contribute to different functional axes, allowing plants to independently modulate different functions by varying network properties at each scale.

 This independence of vein architecture traits within and across spatial scales may enable evolution of multiple alternative leaf network designs with similar functioning.

Introduction

Substantial variation in leaf venation network architecture exists across plant phylogeny (Fig. [1\)](#page-1-0). Some species have a single vein, while others exhibit complex networks with vein density exceed-ing 25 mm mm<sup>-2</sup> (Boyce et al., [2009;](#page-15-0) Brodribb et al., [2010\)](#page-15-0). Some networks are strictly branching, while others comprise thousands of loops with variable shapes. Variation in network architecture also occurs across vein spatial scales (vein orders or sizes) within a single leaf. For example, in many angiosperms, minor veins form loops, whereas major veins only branch (Blonder et al., [2020\)](#page-15-0). While it is still unclear what evolutionary processes have created such a variety of venation networks (Fujita & Mochizuki, [2006\)](#page-15-0), one leading hypothesis is that different networks may reflect selection to optimize different functions (Roth-Nebelsick et al., [2001](#page-16-0); Sack & Scoffoni, [2013;](#page-17-0) Blonder et al., [2020\)](#page-15-0).

Venation networks may contribute to at least six functional axes (hypotheses expanded in Tables [1](#page-2-0), [2\)](#page-3-0): (1) damage resistance to drought (leaf ability to avoid water flow interruption due to xylem conduit implosion or embolism), (2) damage resistance to herbivory (leaf ability to avoid water flow disruption caused by herbivores cutting veins), (3) damage resilience to drought and herbivory (leaf capacity to maintain flow after damages have occurred), (4) flow efficiency (how efficiently water flows through the leaf), and (5) mechanical support (leaf capacity to remain upright in space), which must be traded-off against (6) the leaf construction cost. Due to biophysical and physiological constraints, it might be impossible to construct a network that simul-taneously optimizes all functions (Blonder et al., [2018](#page-15-0), [2020\)](#page-15-0). For instance, it might be difficult to simultaneously achieve high efficiency (via branching networks) and high resilience (via looping networks), as those functions depend on opposing architec-tural features (Roth-Nebelsick et al., [2001](#page-16-0); Ronellenfitsch &

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Katifori, [2019\)](#page-16-0). Depending on the selective forces under which plants have evolved, they may have developed different network architectures as compromises among multiple functions.

While it is known that venation architecture traits (Table [3](#page-6-0)) vary across spatial scales (Blonder et al., [2020\)](#page-15-0), how features at each scale influence different functions remains only partially understood. Some studies propose a functional independence (aka 'labor sharing' Kawai & Okada, [2016,](#page-16-0) [2018\)](#page-16-0) across spatial scales, where features at different vein orders contribute to differ-ent functions (Roth-Nebelsick et al., [2001](#page-16-0); Ueno et al., [2006](#page-17-0); Sack & Scoffoni, [2013\)](#page-17-0). That is, architecture traits at different vein sizes are uncorrelated, allowing plants to independently regulate different functions at each scale. For example, minor vein density may regulate flow efficiency, while major vein density may influence mechanical properties (Sack & Scoffoni, [2013](#page-17-0); Kawai & Okada, [2016](#page-16-0)). By contrast, there could be an integration across vein orders, where venation traits at different spatial

scales are coselected and coordinated to support overall leaf functioning, for example via consistent vein tapering ratios (Price et al., [2007;](#page-16-0) Savage et al., [2010;](#page-17-0) Ronellenfitsch et al., [2015\)](#page-16-0). Those scenarios are not mutually exclusive, and it is possible that the degree of integration vs independence among architecture traits vary across species. Most studies examining those trade-offs were limited to a few species, so we still do not know how the architecture-function space is partitioned among phylogenetically distinct taxa with a wide range of network architectures.

Features not related to the venation architecture can also influence leaf functions and may covary, cancel out, or reinforce the architecture-function trade-offs (Table [2\)](#page-3-0). For instance, outsidexylem conductance can be equally or even more important than the vein-mediated transport to determine flow efficiency (Scoffoni et al., [2023\)](#page-17-0). Similarly, chemical defenses may contribute more to herbivory resistance (Agrawal & Fishbein, [2006\)](#page-15-0), than the mechanical defenses provided by the lignified network Abbreviation Meaning

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(Niklas, [1999](#page-16-0); Kitajima & Poorter, [2010](#page-16-0)). Quantifying the contribution of network architecture features to different leaf functions is therefore essential to understand the constraints that influenced the evolution of diverse leaf networks.

Due to the difficulty in extracting whole-leaf venation net-works (but see Xu et al., [2021\)](#page-17-0) and in obtaining trait datasets describing all functional axes, many hypotheses relating leaf architecture and function (Fig. [2](#page-7-0)) remain to be tested. Previous studies have shown how network architecture features maximize

a single (e.g. Brodribb et al., [2007](#page-15-0), [2016](#page-15-0); Sack et al., [2008;](#page-16-0) Kati-fori et al., [2010](#page-16-0); McKown et al., 2010) or a few functional axes (Walls, [2011;](#page-17-0) Li et al., [2015;](#page-16-0) Kawai & Okada, [2016,](#page-16-0) [2018,](#page-16-0) [2020;](#page-16-0) Blonder et al., [2018;](#page-15-0) Hua et al., [2020;](#page-16-0) Xiong & Flexas, [2022\)](#page-17-0), but we still lack a more integrative picture of multidimensional trade-offs. Rather than maximizing individual functions, venation networks might be constructed to optimize overall leaf functioning. This perspective is consistent with the evolutionary hypothesis of 'multiple alternative designs' in which many phenotypes can achieve similar functioning (Wainwright, [2005;](#page-17-0) Marks & Lechowicz, [2006](#page-16-0)).

Distinguishing between these scenarios can only be assessed by simultaneously investigating multiple functional and architecture axes. In a previous study (Blonder  $et$  al., [2020\)](#page-15-0), we evaluated multifunctional trade-offs across vein spatial scales and found surprisingly weak relationships between venation architecture and leaf functions. However, it remained unclear if those weak trade-offs were caused by a stronger functional role of nonvenation traits or by methodological limitations. The prior study had restricted phylogenetic coverage (only 47 families, all angiosperms, all sampled from one region), incomplete networks missing major veins (networks were extracted from leaf subsections up to  $2 \text{ cm}^2$ ), and insufficient trait dataset to describe all functional axes.

Here, we measured and imputed (16.8% of imputed values) leaf architecture and functional traits (Table 1) from a phylogenetically diverse set of 122 species (Supporting Information Fig. [S1](#page-17-0); Table [S1\)](#page-17-0). Our functional dataset included 13 traits (Table [2\)](#page-3-0) and provided a more complete description of all functional axes. Our venation dataset comprised over 5 million vein segments between 10 and 500 μm of diameter, mostly (95%) extracted from whole leaves. This dataset supported an accurate representation of how key vein features (Blonder et al., [2018,](#page-15-0) [2020\)](#page-15-0) – vein density, vein ramification (branching vs looping), and loop geometry (circular vs elongated) (Table [3\)](#page-6-0) – vary at different spatial scales within a leaf and across species. By combining those two datasets, we were able to (1) describe leaf architecture-function trade-offs, and how they vary across plant clades and vein spatial scales; (2) provide an upper bound estimate of the contribution of venation architecture to each functional axis; (3) identify which venation architecture traits at which scale (small, medium, and large vein width) predict each leaf function; and (4) determine how venation architecture traits across spatial scales interact (via independence and/or integration) to regulate leaf function. Several hypotheses have been previously advanced in the literature for the relationships within and among architecture and functional traits (Table [2\)](#page-3-0). Our dataset allowed a robust assessment of those specific hypotheses (Table [2;](#page-3-0) Fig. [2](#page-7-0)) that have previously been tested with smaller datasets or not at all.

## Materials and Methods

Below, we provide a brief description of how architecture and functional traits were measured. For a detailed description, see Methods [S1](#page-17-0).

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## Species sampling

We sampled 122 species (Table [S1](#page-17-0)) from the living collections of the University of California Botanical Garden at Berkeley (37.87°N, 122.23°W; CA, USA). Our samples included woody and herbaceous species, with origins from all continents except Antarctica, but growing under similar conditions as in a 'com-mon garden' experiment (Perez et al., [2019\)](#page-16-0). For each species, branches (> 1 m long, woody species) or whole plants (herbaceous species) were sampled from 1 to 5 mature individuals, re-cut under water, re-hydrated overnight, and then used for the measurement of functional (Table [2\)](#page-3-0) and architecture traits (Table [3\)](#page-6-0). As most measurements were destructive, different leaves were used for each trait. This sampling approach where the number of replicates within species is reduced to achieve a broader phylogenetic coverage allowed us to investigate trait variation at higher taxonomic levels (clades and families), but had less power at lower levels (genus and species).

## Leaf functional traits

Flow efficiency was quantified as the maximum leaf hydraulic conductance ( $K_{\text{leaf}_{\text{max}}}$ , mmol m<sup>-2</sup> s<sup>-1</sup> MPa<sup>-1</sup>).  $K_{\text{leaf}_{\text{max}}}$  was measured on 4–10 leaves per species using the evaporative flux method (EFM, Sack & Scoffoni, [2012](#page-16-0)) with a pressure-drop flow meter (Melcher et al., [2012](#page-16-0)). EFM more closely approximates the transpiration flow of an *in vivo* leaf (Sack & Scoffoni, [2012](#page-16-0)), as it involves evaporating water out of the lamina while determining the flow rate into the petiole with a flow meter (model PX26- 001GV; Omega Engineering, Norwalk, CT, USA) and the water potential drop across the leaf with a pressure chamber (model 1505D; PMS, Albany, OR, USA).

Damage resistance to drought was measured as the leaf water potentials inducing 50% ( $P_{50}$ , MPa) and 88% ( $P_{88}$ , MPa) loss of  $K_{\text{leaf}_{\text{max}}}$  using the EFM (Sack & Scoffoni, [2012](#page-16-0); Scoffoni et al., [2012](#page-17-0)). To describe xylem resistance to implosion, we also measured xylem conduits' implosion safety index (ISI, dimensionless) from 6 to 8 leaf cross-sectional anatomical images per species, including images of the petiole, major and minor veins (Matos et al., [2024](#page-16-0)). For each image, xylem cell wall maximum diameter  $(b)$  and double intervessel cell wall thickness  $(t)$  were manually measured on all or, at most, 10 adjacent xylem conduits using the software IMAGEJ (<https://imagej.nih.gov/>). ISI was calculated as t/b (Hacke et al., [2001](#page-16-0); Blackman et al., [2010\)](#page-15-0).

Damage resistance to herbivory was quantified as the specific work to punch (SWP, kJ m<sup>-2</sup> m<sup>-1</sup>) and to shear (SWS, J m<sup>-2</sup>) a leaf (3–4 leaves per species) using a universal testing machine (UTM, Test stand ES30 and force gauge series M5; Mark-10, Copiague, NY, USA). Punching tests involved forcing a rod of known cross-sectional area through the leaf midrib or lamina, whereas shearing tests involved using a single blade to make a transverse cut across the leaf midrib and lamina (Read et al., [2005](#page-16-0); Onoda et al., [2011\)](#page-16-0). We also quantified the total phenol content (Phe,  $g g^{-1}$ ) in dried leaves using the Folin–Ciocalteu assay (Ainsworth & Gillespie, [2007](#page-15-0)). Phe can be used as a partial proxy of chemical defense against herbivores (Matsuki, [1996\)](#page-16-0).

#### <span id="page-6-0"></span>Table 3 Leaf venation architecture traits (abbreviation, unit and definition).





The example illustrations show networks with high or low values of each venation architecture traits for the small veins only (yellow lines), while features of medium (orange) and large (purple) veins remain constant. In reality, variation in architecture traits can occur across all scales (small, medium, and large veins).

Damage resilience was quantified as the average change in leaf hydraulic conductance ( $\Delta K_{\text{leaf}_{\text{mean}}}$ , %) after the lamina or the midrib(s) were cut at one-third distance from the leaf base (Dela-ney & Higley, [2006](#page-15-0); Sack et al., [2008\)](#page-16-0). Severing treatments subjected leaves to both physical/herbivory (because of the cut) and hydraulic/drought (because air could enter through the cut con-duits) damages (Sack et al., [2008;](#page-16-0) Peschiutta et al., [2016\)](#page-16-0). Forty-eight hours after treatment, we excised the damaged leaves and measured  $K_{\text{leaf}}$  using the EFM.  $\Delta K_{\text{leaf}}$  was calculated separately for the midrib and lamina, that is  $\Delta K_{\text{leaf}} = [(\Delta K_{\text{leaf} \text{midib}} \text{ or }$  $\Delta K_{\text{leaf}_\text{lamina}} \times 100$ )/ $K_{\text{leaf}_\text{max}}$ ] – 100, and then averaged for each sample to obtain  $\Delta K_{\text{leaf}_{\text{mean}}}$ .  $\Delta K_{\text{leaf}_{\text{mean}}}$  < 0 indicates lower resilience (i.e.  $K_{\text{leaf}}$  declines after damage), while  $\Delta K_{\text{leaf}_{\text{mean}}} \ge 0$  indicates higher resilience.

Mechanical support was described by the leaf flexural modulus of elasticity ( $\varepsilon$ , MN m<sup>-2</sup>; Read *et al.*, [2005\)](#page-16-0), measured on 3-4 leaves per species using the UTM. For each leaf, 3-point bending tests were conducted twice, to obtain  $\varepsilon$  for the whole leaf ( $\varepsilon_{\text{whole}}$ ), and leaf lamina ( $\varepsilon_{\text{lamina}}$ ). Leaves were placed in the UTM with their longitudinal axis parallel to the bending fixture.

Construction cost was described as the leaf mass per area (LMA,  $\rm g$  m $^{-2}$ ), measured in 3–5 mature leaves per species. Leaves were first scanned to obtain leaf area, and then oven-dried at 50°C for 48 h to determine their dry mass (B2-Series; VWR, Radnor, PA, USA). LMA was then calculated as leaf dry mass/leaf area (Pérez-Harguindeguy et al., [2016\)](#page-16-0).

ISI, LMA, SWP<sub>lamina</sub> and  $K_{\text{leaf}_{\text{max}}}$  were measured in all 122 species. For various reasons (Methods [S1](#page-17-0)), missing data occurred for the other functional traits, with variable percentages of missing values ranging from 1% (Phe) to 74% ( $P_{50}$  and  $P_{88}$ ). Therefore, caution must be used when interpreting the results for those last two traits.

#### Leaf venation architecture traits

To obtain the architecture traits, leaves were pressed flat, dried, and chemically cleared and stained to highlight veins (Blonder et al., [2018;](#page-15-0) Pérez-Harguindeguy et al., [2016](#page-16-0)). Next, cleared leaf samples were imaged using a 100-mm macro-objective lens (Tokina, Huntington Beach, CA, USA) and digital camera (EOS 6D; Canon, Southend-on-Sea, UK) or a transilluminated scanner (Epson Perfection V850 Pro, Los Alamitos, CA, US), producing high-resolution images  $(47-144 \text{ pixels mm}^{-1})$  of the whole leaf. This process resolved veins  $\geq 10$  µm in diameter. Therefore, all larger veins were accurately imaged in all species, but in species with minor venation smaller than this threshold, the architecture traits at smaller spatial scales missed some veins.

Leaf images were processed using the LEAFVEINCNN software, v.1.3 and newer (Xu et al., [2021](#page-17-0); software and manual available from doi: [10.5281/zenodo.4007731\)](https://doi.org/10.5281/zenodo.4007731). LeafVeinCNN relies on an ensemble of three convolutional neural networks to automatically

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## (a) Hypothetical combination of leaf venation architecture traits

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Fig. 2 Hypothesized combinations of leaf venation architecture traits (ER, loop elongation ratio; MST, minimum spanning tree ratio; VD, vein density) at three vein spatial scales (small, medium, and large) that would have evolved if each leaf functional axis was independently optimized: (a) Radar chart of hypothesized trait combinations. (b) Damage resistance to drought should be higher in networks with lower density of large veins (low VD<sub>large</sub>), branching large and medium veins (low  $MST<sub>large</sub>$  and  $MST<sub>medium</sub>$ ), and more circular loops in small veins (low ERsmall); (c) Damage resistance to herbivory should be higher in networks with higher large vein density (high VD<sub>large</sub>) and more circular loops at all scales (low ER); (d) Damage resilience to drought and herbivory should be higher in networks with higher density of large and small veins (high VD<sub>small</sub> and VD<sub>large</sub>), palmate venation (more than one midrib), and more loops (low MST) at all scales; (e) Flow efficiency should be higher in networks with higher density (high VD<sub>small</sub>) of branching (high MST<sub>small</sub>) small veins; (f) Mechanical support should be higher in networks with higher large vein density (high VDlarge) and more loops in small veins (low  $MST<sub>small</sub>$ ; and (g) Construction cost should be higher in networks with higher density (high VDlarge and VDmedium) of large and medium veins.

segment veins and produce a spatial graph representation of the networks. The program also uses hierarchical loop decomposition algorithms (Katifori & Magnasco, [2012](#page-16-0)) to extract multiscale venation statistics, which describe how venation architecture traits vary across spatial scales (Blonder  $et$   $al$ ,  $2020$ ), that is across each value of vein width  $(r_{\min})$ . We focused our analysis on three statistics–vein density (VD), minimum spanning tree ratio (MST), and loop elongation ratio (ER) (Table  $3$ ) – previously identified as key traits to describe venation architecture (Blonder et al., [2018,](#page-15-0) [2020\)](#page-15-0).

Extensive quality control steps, including additional hand-tracing in some samples, followed standard protocols, ensuring the accuracy of the segmented networks and the multiscale statistics. Architecture traits could not be obtained for two species (Aucuba japonica Thunb. and Nymphaea sp.) because chemical treatments failed to properly digest nonvenation tissues. In seven monocot species, leaves were too long ( $>$  30 cm) to be fully processed, so a leaf segment representing  $c$ . 20–50% of the total leaf area was analyzed.

#### Statistical analysis

To evaluate whether species in different clades and/or veins at different sizes occupy different portions of the architecture-function space (Goal 1), we carried out a principal component analysis (PCA) with all traits. Before the PCA, we imputed missing values (Methods [S1\)](#page-17-0) using a Bayesian hierarchical probabilistic matrix factorization (BHPMF; Schrodt et al., [2015\)](#page-17-0). BHPMF imputes values based on the taxonomic hierarchy and correlation structure within the trait matrix. Briefly, we  $log_{10}$ - and *z*-transformed all trait values, performed 50 imputations (Table [S2\)](#page-17-0), filtered out implausible values, and then calculated the mean and SD of the imputed values. To validate the imputation, we regressed original vs imputed values and evaluated the  $\mathbb{R}^2$  (Methods [S1\)](#page-17-0). Next, venation traits (VD, ER, and MST) were binned into 50  $r_{\text{min}}$ bins, spanning 0.01 mm ( $r_{\min}$  < 0.01 mm, veins too small to be distinguishable) to  $0.5$  mm ( $r_{\text{min}} > 0.5$  mm, too few veins sampled, see Fig. [S2](#page-17-0)). Finally, trait values were centered and scaled (z-transformed) to improve comparability and reduce bias

toward traits with higher variance. VD and ER were also squareroot-transformed to improve normality. The broken-stick method was used to determine the principal components to be retained. The retained principal components were visualized using 95% confidence ellipses at each clade and at each  $r_{\min}$ .

To quantify the contribution of architecture traits to each function (Goal 2), we fitted gradient boosting machine (GBM) models. GBM is a machine-learning ensemble method that effectively captures complex nonlinear interactions between predictor variables (Natekin & Knoll, [2013\)](#page-16-0). We fitted our models using one functional trait at a time as the response variable, and clade plus architecture traits as predictor variables. To make an interpretable assessment of the contribution of architecture traits at different spatial scales, we binned VD, MST, and ER at three scales to represent small, medium, and large veins. Because the range of vein sizes vary across leaves, we used two complementary approaches (scaled and unscaled  $r_{\text{min}}$ ) to classify veins into size categories (Methods [S1](#page-17-0)), allowing us to investigate how both relative (scaled  $r_{\text{min}}$ ) and absolute veins sizes (unscaled  $r_{\text{min}}$ ) influence architecture-function trade-offs. To fit each GBM, we split data 80%/20% between training and test sets. Then, we used the  $h2o.$ *automl* function as implemented in the  $h2o$  R-package to perform a hyperparameter search over the GBM parameters. To prevent model overfitting, hyperparameter tuning was done with a maximum running time of 30 s and a threefold cross-validation. Model performance was assessed using root-mean-square error (RMSE), and the best model (lowest RMSE) for each functional trait was selected using the function h2o.get\_best\_model. From each best model, we obtained the total variance explained as an estimate of the contribution of venation architecture traits to each functional axis (Goal 2). Due to the lower sample size  $(n = 32)$ , GBM models for  $P_{50}$  and  $P_{88}$  were fitted using both the complete-case and the BHPMF-imputed datasets. To account for the uncertainty around the imputed values, we ran 50 GBM models for  $P_{50}$  and  $P_{88}$ , using imputed values obtained in each of the 50 imputations. We report the GBM results averaged across those iterations.

To identify which architecture traits at which scale predict leaf functions (Goal 3), we obtained the influence value of each predictor variable in each best GBM model. Variable influence was obtained using both permutation importance and SHapley Additive exPlanation (SHAP) values (Štrumbelj & Kononenko, [2014](#page-17-0)). Variable importance, ranging from 0 (lowest importance) to 1 (highest importance), was determined using the *h2o.varimp* function, which measures the increase in the model RMSE after variable values are permuted. To compare the importance of predictor variables between the best model and the other GBM models fitted with the same data, we used the function h2o.varimp\_heatmap, which produces a heatmap of variable importance across multiple GBM models. SHAP values were determined using h2o.predict\_contributions function, and they measure the impact of every predictor variable on the model prediction for each instance of the data. Thus, predictor variables with larger absolute SHAP values have a larger contribution to explain the response variable.

To determine how architecture traits across scales interact to predict leaf functions (Goal 4), we measure the strength of pairwise interactions using the H-statistic (Friedman &

Popescu, [2008\)](#page-15-0), as implemented in *Interaction\$new* from the *iml* R-package. H-statistics measure how much of the variation of the predicted outcome depends on a given pairwise interaction and vary from 0 (no interaction) to 1 (100% of variance is due to interactions).

To test for differences in functional traits across plant clades, and to test our hypothesis H3c (Table [2](#page-3-0)) that resilience varies across venation types (parallel, palmate, and pinnate), we used phylogenetic ANOVA tests followed by post hoc pairwise tests with Benjamini & Hochberg ([1995](#page-15-0)) P-value adjustments. Phylogenetic ANOVA tests were implemented following Garland Jr. et al. [\(1993](#page-15-0)) using the function *phyloANOVA* from the PHYTOOLS R-package. All analyses were carried out using R v.4.3.1 (R Core Team, [2023\)](#page-16-0). R code to reproduce all analyses is available at: [https://github.com/ilamatos/venation\\_tradeoffs/](https://github.com/ilamatos/venation_tradeoffs/).

## **Results**

#### Leaf functional and architecture traits vary across species and scales

We found substantial variation in functional traits across species (Tables [S1](#page-17-0)–S3). For example, LMA varied c. 15-fold across species (from 18.91 to  $276.20 \text{ g m}^{-2}$ ), while  $K_{\text{leaf}_{\text{max}}}$  varied c. 232-fold (from 0.15 to 34.83 mmol  $\text{m}^{-2} \text{ s}^{-1} \text{ MPa}^{-1}$ ). SWP<sub>midrib</sub> was on average  $(226.08 \text{ kJ m}^{-2} \text{ m}^{-1})$  c. 2x higher than  $SWP_{\text{lamina}}$  (128.83 kJ m<sup>-2</sup> m<sup>-1</sup>). By contrast, SWS<sub>lamina</sub> (6309 J m<sup>-2</sup>) and  $\varepsilon_{\text{lamina}}$  (78.77 MN m<sup>-2</sup>) were on average c. 3× higher than SWS<sub>midrib</sub> (2316 J m<sup>-2</sup>) and  $\varepsilon_{\rm whole}$  (28.66 MN  $m^{-2}$ ). When leaf midrib(s) and lamina were damaged,  $K_{\text{leaf}_{\text{max}}}$ decreased on average by c. 55%. However, some species ( $n = 11$ ) experienced an increase in  $\Delta K_{\text{leaf}_{\text{mean}}}$  (from 0.07% up to c. 295%). When we applied phylogenetic ANOVA tests to compare functional traits across clades, we found significant differences only in SWS<sub>lamina</sub>, which was lower in basal eudicot species (Fig. [S3](#page-17-0)).

Architecture traits demonstrated high variation among species and clades (Figs [S4](#page-17-0)–[S6](#page-17-0)), and complex patterns across vein sizes. As  $r_{\text{min}}$  increased, VD decreased (Fig. [S4\)](#page-17-0) and MST increased (Fig.  $S$ 5), whereas ER peaked at intermediate  $r_{\min}$  values (Fig. [S6](#page-17-0)), reflecting a tendency of more veins and loops at smaller spatial scales, and more elongated loops at medium-sized veins.

#### Leaf architecture-function trade-offs are in general weak

Principal component analysis identified three significant axes (Fig. [S5](#page-17-0)), which cumulatively explained 44% of the total variation (Figs [3,](#page-9-0) [S8\)](#page-17-0), suggesting weak generalized architecturefunction trade-offs across vein scales and plant phylogeny. Despite the fundamental differences in drought- and herbivoryinduced damages (the former can spread across the network while the latter may not), resistance to both factors covaried along the first principal component (PC1, Fig. [3](#page-9-0)). Leaves on the left PC1 side have high resistance against both herbivory (higher SWP) and drought (more negative  $P_{88}$ ), while leaves on the right PC1 side have low resistance (Tables [S4,](#page-17-0) [S5](#page-17-0)). Architecture

<span id="page-9-0"></span>

Fig. 3 First (PC1), second (PC2), and third (PC3) principal components of leaf venation architecture (VD, MST, ER, colored in red) and functional traits ( $K_{\text{leaf}_{\text{max}}}$ ,  $P_{50}$ ,  $P_{88}$ , ISI, SWP<sub>midrib</sub>, SWP<sub>lamina</sub>, SWS<sub>midrib</sub>, SWS<sub>lamina</sub>, ΔK<sub>leafmean</sub>, ε<sub>whole</sub>,  $\varepsilon$ <sub>lamina</sub>, LMA, Phe, colored in black) across 50 bins of vein width sizes ( $r_{\text{min}}$ ). In panels (a, c, e, g), the 95% confidence ellipses enclose the data at each plant phylogenetic clade (ferns, basal angiosperms, monocots, basal eudicots, rosids, and asterids), or at each vein spatial scale  $(r_{\text{min}})$ . Panels (b, d, f, h) show the principal component analysis loadings. Note that leaf architecture traits (shown in red) vary across vein sizes  $(r_{\text{min}})$ , while functional traits (shown in black) do not. The main loadings for PC1, PC2, and PC3 are shown in bold. Other trait loadings are numbered from 1 to 9. Parenthetical values indicate the percentage variance explained by the three first principal component axes. Trait abbreviations: ER, elongation ratio; ISI, Implosion safety index;  $K_{\text{leaf}_{\text{max}}}$ , maximum leaf hydraulic conductance; LMA, leaf mass per area; MST, minimum spanning tree ratio;  $P_{50}$ , leaf water potential inducing 50% loss of  $K_{\text{leaf}_{\text{max}}}$ ;  $P_{88}$ , leaf water potential inducing 88% loss of  $K_{\text{leaf}_\text{max}}$ ; Phe, Phenol content; SWP\_L, specific work to punch the leaf lamina; SWP\_M, specific work to punch the midrib; SWS\_L, specific work to shear the lamina; SWS\_M, specific work to shear the midrib; VD, vein density;  $\Delta K_{\text{leaf}}$ , mean change in  $K_{\mathrm{leaf}_{\mathrm{max}}}$  after damaging the leaf lamina and midrib;  $\varepsilon$ \_L, leaf flexural modulus of elasticity for the leaf lamina;  $\varepsilon_N$ W, leaf flexural modulus of elasticity for the whole leaf.

traits had a low contribution to PC1 (Table [S4](#page-17-0)). PC2 reflected a network architecture-support axis (Fig. 3). Leaves at the upper PC2 axis are more elastic (lower  $\varepsilon$ ) and have a high density (higher VD) of small veins (lower  $r_{\min}$ ) forming circular loops (lower ER and MST), while leaves at the bottom PC2 axis show an opposite combination of traits. PC3 further reflected the trade-off between VD and MST (Fig. [S8\)](#page-17-0). Fern species spread along PC1, reflecting their broad range of drought and herbivory resistances, whereas monocots spread along PC2 reflecting the existence of different venation architectures (e.g. parallel and palmate) within this clade (Fig. 3). Vein sizes  $(r_{\min})$  differentiated

along PC2 (Fig. 3) and PC3 (Fig. [S8](#page-17-0)), reflecting the tendency of small veins (lower  $r_{\text{min}}$ ) to have higher density and more circular loops.

#### Venation architecture traits contribute to leaf support, cost, and resistance

Architecture traits plus clade explained half or more of the total variance (Fig. [4](#page-10-0)) for traits describing mechanical support, cost, and resistance but had a lower explanatory power for determining leaf resilience and efficiency. Except for Phe and SWS<sub>midrib</sub>, the

<span id="page-10-0"></span>

Fig. 4 Results of gradient boosting machine (GBM) assessing variation in architecture (predictor variables) and functional (response variables) tradeoffs across vein spatial scales (small, medium, and large veins) and plant phylogeny (clade). The total percentage of variance explained by the best GBM model indicates how much the variance in venation architecture traits across species contribute to explain the variance in 13 leaf functional traits ( $P_{50}$ ,  $P_{88}$ , ISI, SWP<sub>midrib</sub>, SWP<sub>lamina</sub>, SWS<sub>midrib</sub>, SWP<sub>lamina</sub>, ΔK<sub>leafmean</sub>,  $K_{\text{leaf}_{\text{max}}}$ ,  $\varepsilon_{\text{whole}}$ ,  $\varepsilon_{\text{laminar}}$ , LMA, Phe) describing six functional axes (resistance to drought, resistance to herbivore, resilience, flow efficiency, mechanical support, and construction cost). All models were run with complete-case dataset and either with scaled (scaled  $r_{min}$ ) or unscaled (unscaled  $r_{min}$ ) vein sizes. Trait abbreviations: ISI, Implosion safety index;  $K_{\text{leafmax}}$ , maximum leaf hydraulic conductance; LMA, leaf mass per area; P<sub>50</sub>, leaf water potential inducing 50% loss of  $K_{\text{leafmax}}$ ;  $P_{88}$ , leaf water potential inducing 88% loss of  $K_{\text{leaf}_{\text{max}}}$ ; Phe, Phenol content; SWP<sub>lamina</sub>, specific work to punch the leaf lamina; SWP<sub>midrib</sub>, specific work to punch the midrib; SWS<sub>lamina</sub>, specific work to shear the lamina; SWS<sub>midrib</sub>, specific work to shear the midrib;  $\Delta K_{\text{leaf}_{\text{mean}}}$ , mean change in  $K_{\text{leaf}_{\text{max}}}$  after damaging the leaf lamina and midrib;  $\varepsilon_{\text{lamina}}$ , leaf flexural modulus of elasticity for the leaf lamina;  $\varepsilon_{\text{whole}}$ , leaf flexural modulus of elasticity for the whole leaf.

variance explained was similar in models using absolute (unscaled  $r_{\text{min}}$ ) and relative (scaled  $r_{\text{min}}$ ) vein sizes. For  $P_{50}$  and  $P_{88}$ , variance explained was higher in models ran with complete-case (Fig. 4), than with imputed trait values  $(R^2: P_{50_{\text{scaled}}} = 0.30;$  $P_{50_{\rm unscaled}} = 0.32; P_{88_{\rm scaled}} = 0.23; P_{50_{\rm unscaled}} = 0.62$ ). Tables [S6](#page-17-0) and [S7](#page-17-0) show hyperparameters values for the best selected models and performance metrics for all GBM models, respectively.

#### Venation architecture contribution to leaf function vary across spatial scales

Architecture traits at different spatial scales contributed to different leaf functions (Fig. [S9\)](#page-17-0). Overall, clade was not ranked as an important variable to explain variation in most of the functional traits, suggesting that architecture-function trade-offs vary more within rather than between clades. The order of predictor variables' importance changed between GBM models based on the absolute or relative vein sizes (Fig. [S9](#page-17-0)), and between different GBM models fitted using the same data (Figs [S10](#page-17-0), [S11\)](#page-17-0). But, in most cases, the direction of the predictor variables' effect on the response variable was similar across different models.

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The impact of each predictor variable on each functional trait is displayed in the SHAP summary plots (Figs [5](#page-11-0), [S12, S13\)](#page-17-0). In the SHAP plot, each point represents an instance of data (i.e. one species). The point color indicates the normalized value of the corresponding predictor variable, with pink indicating high values and blue indicating low values. The  $\nu$ -axis shows the top five (from top to bottom) predictor variables most important to explain the response variable. The x-axis represents the SHAP contribution values. SHAP signal indicates whether the predictor variable increases (SHAP  $> 0$ ) or decreases (SHAP  $< 0$ ) the response variable, while SHAP magnitude measures how strong this positive or negative impact is. Thus, a SHAP value around zero indicates that the predictor variable has an irrelevant effect on the response variable, while a larger absolute SHAP value indicates a greater effect. For example, in Fig. [5\(c\)](#page-11-0), loop elongation in small veins ( $ER_{small}$ ) followed by  $MST_{medium}$  and  $MST_{large}$ were the three most important variables to explain interspecific variation in leaf resistance to herbivory described as SWP<sub>lamina</sub>. Species with more elongated loops in small veins (i.e. higher ERsmall indicated by the pink-colored points on the right side of the  $ER_{small}$  row) tend to exhibit higher resistance to herbivory (higher SWP<sub>lamina</sub> indicated by the more positive SHAP values). Additionally, species with more loops on medium size veins (i.e. lower MST<sub>medium</sub> indicated by the blue-colored points) are associated with higher SWP<sub>lamina</sub> values (more positive SHAP values). By contrast, species with more loops in large size veins (i.e. lower  $MST_{large}$ ) are associated with lower SWP<sub>lamina</sub> (more negative SHAP values). Those results only partially support our hypothesis (H2), as we found no strong evidence that higher density of large veins contributes to higher resistance to herbivory (Figs [S12](#page-17-0), [S13](#page-17-0)). Phe was associated with networks of higher vein density and more circular loops at all scales (Figs [S12](#page-17-0), [S13](#page-17-0)). Contrary to H2c, we found no evidence that high investment in secondary chemistry offsets investment in physical defense provided by the lignified network.

Contrary to our expectations (H1a and H1c), resistance to drought was higher in networks with higher density of larger veins and with intermediary values of ER (Figs [5a](#page-11-0), [S12](#page-17-0), [S13](#page-17-0)). As expected (H1b), more tree-like small veins resulted in higher resistance to drought, but this effect was weak. Because of the greater uncertainty associated with the imputed  $P_{50}$  and  $P_{88}$ values (Table [S2](#page-17-0)), when we averaged the results across the 50 GBM imputed-based models, we basically found no clear effect of venation architecture traits on drought resistance (Fig. [S14\)](#page-17-0).

Resilience was higher in networks with higher density of large veins and either high or low density of small veins, thus partially supporting H3a (Figs [5d,](#page-11-0) [S12,](#page-17-0) [S13\)](#page-17-0).  $\Delta K_{\rm leaf_{mean}}$  was also higher in networks with intermediate values of ER<sub>medium</sub> and ER<sub>large</sub>, more loops in large veins, and more branching minor veins, thus partially supporting H3b. Contrary to H3c, we found no significant differences in  $\Delta K_{\text{leaf}_{\text{mean}}}$  ( $F = 0.2322$ , P-value = 0.906),  $\Delta K_{\text{leaf}_{\text{midrib}}}$  $(F = 0.3716,$  P-value = 0.841), or  $\Delta K_{\text{leaf}_\text{lamin}}$   $(F = 0.0802,$  $P$ -value = 0.968), between palmate, pinnate, and parallel leaves.

As hypothesized (H4a), flow efficiency was higher in networks with higher density of small veins (Figs [5b,](#page-11-0) [S12](#page-17-0), [S13](#page-17-0)). However, higher  $K_{\text{leaf}_{\text{max}}}$  was found in networks with intermediate, rather

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Fig. 5 SHapley Additive exPlanations (SHAP) summary plots for six leaf functional axes. SHAP values measure the impact of predictor variables (i.e. plant clades and vein density (VD), loop elongation ratio (ER), and minimum spanning tree ratio (MST) at three spatial scales – large, medium, and small) on the response variables (leaf functional traits) considering the interactions between predictors. (a) Resistance to drought –  $P_{50}$  (leaf water potential inducing 50% decline in leaf hydraulic conductance), (b) Efficiency –  $K_{\text{leaf}_{\text{max}}}$  (maximum leaf hydraulic conductance), (c) Resistance to herbivory – SWP $_{\text{lamina}}$  (specific work to shear leaf lamina), (d) Resilience –  $\Delta K_{\text{leaf}_{\text{mean}}}$  (percentage decline in leaf hydraulic conductance after damage), (e) Mechanical support –  $\varepsilon$ <sub>lamina</sub> (leaf flexural modulus of elasticity for the leaf lamina), (f) Construction cost – leaf mass per area. Note that this figure only presents SHAP values for the top five most important predictor variables, and for just one functional trait per functional axis. Complete SHAP summary plots for all 13 functional traits measured in this study are shown in Fig. [S11](#page-17-0). All results present here were run with complete-case dataset and unscaled  $r_{\text{min}}$  (i.e. absolute vein sizes), for the scaled  $r_{\text{min}}$  (i.e. relative vein sizes) results, see Fig. [S12](#page-17-0). For traits abbreviations, see Table [1](#page-2-0).

than low MST<sub>small</sub> and more elongated loops in medium and large veins.

Mechanical support was higher in networks with more looping small veins and higher large vein density, although in some species high VD<sub>large</sub> was associated with lower  $\varepsilon_{\text{whole}}$ , thus only partially supporting H5a. The presence of elongated loops in medium veins was also important to explain  $\varepsilon$  in our dataset (Figs 5e, [S12](#page-17-0), [S13](#page-17-0)). Note that those relationships were highly influenced by a few monocot species with parallel venation and high elasticity. As expected (H6a), construction cost was higher in leaf networks with higher vein density, but this effect was not particularly strong for large veins (Figs 5f, [S12,](#page-17-0) [S13\)](#page-17-0). Fig. [6](#page-14-0) summarizes the results for the architecture-function trade-offs across the six leaf functional axes.

#### Venation architecture traits across spatial scales interact weakly to regulate leaf function

Overall, pairwise interactions between predictor variables were weak (Fig.  $S15$ ), with *H*-statistic < 0.3 (Table [S8\)](#page-17-0), suggesting low integration across architecture traits at different vein orders.

Fig. [S15](#page-17-0) shows the strongest interactions in our dataset for the GBM models with scaled and unscaled  $r_{\min}$ , respectively.

#### **Discussion**

Overall, we found that: (1) generalized leaf architecture-function trade-offs across vein spatial scales and plant phylogeny are weak; (2) architecture traits play a stronger role on leaf mechanical support and damage resistance, than on flow efficiency and damage resilience; (3) architecture traits at different spatial scales contribute to different leaf functions, which potentially reduces the necessity of trade-offs among functional axes driven by venation architecture; and (4) pairwise interactions between architecture traits both at a single scale (e.g.  $VD_{\text{large}} \times ER_{\text{large}})$  and across spatial scales (e.g.  $VD_{small} \times VD_{large}$ ) are weak, reflecting no widespread trade-offs among different aspects of network architecture.

Those results suggest that given the existence of different constraints (e.g. biophysical, physiological, and phylogenetic) leaf venation networks reflect a compromise among multiple competing functions (Ronellenfitsch et al., [2015](#page-16-0); Patino-Ramirez & Arson, [2020](#page-16-0)). This means that no single optimal venation

network exists, but instead, there might be multiple trait combinations or alternative designs of network architectures (Wainwright, [2005;](#page-17-0) Marks & Lechowicz, [2006](#page-16-0)) that can be deployed to achieve equivalent overall functioning. The independence of venation architecture traits within and across vein spatial scales, reflected by the weak pairwise interactions in our dataset, could allow more freedom for adjusting traits and functioning to the environment (Li et al., [2015;](#page-16-0) Méndez-Alonzo et al., [2019](#page-16-0)). This perspective contrasts with prior conclusions about leaf venation in which it was proposed that single functions were optimized (Sack & Tyree, [2005;](#page-17-0) Zwieniecki et al., [2006](#page-17-0); Noblin et al., [2008](#page-16-0); Deans et al., [2020\)](#page-15-0) and reinforces recent ideas about the high dimensionality of adaptation in plants (Baraloto et al., [2010;](#page-15-0) He et al., [2020\)](#page-16-0). This means that species may achieve similar levels of function in the same environment despite dissimilar venation architectures. Because leaf functions are determined not only by network architecture but also by multiple biochemical, anatomical, morphological, and physiological venation and nonvenation traits operating at different spatial scales (Table [2\)](#page-3-0), there could be numerous trait combinations yielding similar functioning. Each leaf represents the realization of one viable combination among those multiple possibilities. Thus, it is not surprising that we failed to identify strong and widespread architecture-function trade-offs across the plant phylogeny. Nevertheless, we did find clear linkages between venation architecture and leaf functions, particularly for mechanical support, damage resistance, and cost. We discuss how variation in key venation architecture traits across spatial scales influence each leaf functional axis independently, even though leaf networks are likely selected for achieving sufficient performance in each functional axis, rather than an optimum performance in any single function.

#### Influence of venation architecture on each leaf function

Damage resistance to drought: Although larger veins seem to be more prone to embolisms (Brodribb et al., [2016;](#page-15-0) Scoffoni et al., [2017b\)](#page-17-0), we found that networks with a higher density of larger veins may be more resistant to drought. This result was supported by previous empirical studies (Scoffoni et al., [2011](#page-17-0); Nardini et al., [2014](#page-16-0); Xiong & Flexas, [2022](#page-17-0)) and could be explained by the concept of relative leverage between veins (Sack & Scoffoni, [2013\)](#page-17-0). When  $VD<sub>large</sub>$  is higher, large veins have less leverage relative to the small vein system, and embolisms in major veins should cause a smaller decrease in  $K_{\text{leaf}}$ . Conversely, when  $VD_{small}$  is higher, large veins gain leverage, and embolisms in large veins should cause great loss of conductance. Therefore, since embolisms typically begin in the largest veins and progresses into smaller ones (Brodribb et al., [2016](#page-15-0); Scoffoni et al., [2017b\)](#page-17-0), a higher VD<sub>large</sub> is likely more effective for reducing overall leaf hydraulic vulnerability to drought than a higher VD<sub>small</sub> (Sack & Scoffoni, [2013](#page-17-0)). Thus, despite the overall trend of independence among architecture traits, in some circumstances, a coordination in network architecture across spatial scales could be important to maximizing some leaf functions (Kawai & Okada, [2018\)](#page-16-0).

Damage resistance to herbivory was overall higher in networks with small veins forming circular loops (lower ERsmall and  $MST_{small}$ ), corroborating the hypothesis that reticulation in small veins can increase deterrence against chewing/cutting damage by providing more ways to stop the propagation of mechanical frac-tures (Niklas, [1999;](#page-16-0) Fiorin et al., [2016\)](#page-15-0). However, contrary to previous studies, we found no evidence that a higher density of large veins increases damage resistance to herbivory (Sack et al., [2008](#page-16-0); Kitajima & Poorter, [2010](#page-16-0); Kawai & Okada, [2016](#page-16-0)) nor that a high investment in chemical defense can offset investment in mechanical resistance by the lignified network (Blonder et al., [2020;](#page-15-0) Duarte et al., [2023\)](#page-15-0). Importantly, we did not directly measure herbivory attack nor did we quantify other secondary metabolites important for chemical defenses against herbivory, so those hypotheses cannot be completely ruled out.

Mechanical support was largely explained by venation architecture features, and higher in networks with higher density of larger veins and more looping small veins. Those results corroborate previous studies showing that larger and more sclerified veins can increase leaf stiffness (Onoda et al., [2015;](#page-16-0) Kawai & Okada, [2016](#page-16-0)) as they act as beams that support the leaf weight (Niklas, [1999\)](#page-16-0), while more reticulated small veins can increase stiffness by providing reinforcing cross-linkages that resist mechanical loading (Niklas, [1999](#page-16-0); Roth-Nebelsick et al., [2001](#page-16-0)). Importantly, in our dataset, the architecture-function trade-offs along the mechanical support axis were largely influenced by monocot species with parallel venation and stiff leaves. In those species, the parallel arrangement of the large veins, reinforced with small transverse veins, leads to a very efficient stabilization of the leaf against bending forces (Roth-Nebelsick et al., [2001\)](#page-16-0). We also found a novel result that more elongated medium veins increase leaf stiffness, perhaps because they form cross-links between primary and secondary veins, helping to stabilize the whole leaf in the same way that triangular trusses stabilize bridges.

Construction cost was associated with high densities of both small and large veins, indicating that even though minor veins usually occupy a small volume inside the leaf (Sack & Scoffoni, [2013](#page-17-0)), they may still impact the final leaf cost, probably due to their more than fourfold higher tissue density compared with mesophyll and epidermis tissues (Poorter et al., [2009](#page-16-0)). Those results contradict previous studies showing that leaf construction cost and the densities of both major and minor veins form independent axes (Kawai & Okada, [2020](#page-16-0)). Our work, based on a larger and more phylogenetically diverse dataset, suggests that venation architecture features are important in explaining variation in leaf cost across species, and do not form independent axes in the architecture-function space.

Damage resilience: Overall, venation architecture traits had a low contribution to this functional axis. Resilience was slightly higher in networks with a high density of looping large veins. This architecture design may increase resilience by providing more redundant pathways for continued long-distance flow after damages (Sack et al., [2008;](#page-16-0) Katifori et al., [2010\)](#page-16-0). More loops in small veins, however, decreased resilience, probably because too much reticulation in small veins can lead to a faster spread of embolisms (Loepfe et al., [2007](#page-16-0); Mrad et al., [2021\)](#page-16-0). By contrast,

less reticulation may increase resilience by restricting damages to a single conduit or group of conduits (Sack & Scoffoni, [2013](#page-17-0)) and maybe also by facilitating embolism repair (Schenk et al., [2008](#page-17-0)). Therefore, optimal architecture for resilience may occur when networks combine high redundancy (more loops) in major veins with high sectoriality (less loops) in minor veins. In contrast to Sack et al. ([2008\)](#page-16-0), we found no difference in resilience between palmate, pinnate, and parallel leaves. This is probably



Barbacenia purpurea

<span id="page-14-0"></span>Fig. 6 Radar charts, based on observed data, for the combination of leaf venation architecture traits (ER, elongation ratio; MST, minimum spanning tree ration; VD, vein density) at three vein spatial scales (small, medium, large) that maximizes each leaf functional axis independently: (a) Resistance to drought –  $P_{50}P_{50}$  (leaf water potential inducing 50% decline in leaf hydraulic conductance), (b) Resistance to herbivory – SWP<sub>lamina</sub> (specific work to shear leaf lamina), (c) Resilience –  $\Delta K_{\text{leaf}_\text{mean}}$  (percentage decline in leaf hydraulic conductance after damage), (d) Efficiency –  $K_{\text{leaf}_\text{max}}$  (maximum leaf hydraulic conductance), (e) Mechanical support – ε<sub>lamina</sub> (leaf flexural modulus of elasticity for the leaf lamina), (f) Construction cost – leaf mass per area. In the radar charts, point sizes indicate the predictor variable importance, with bigger points reflecting higher importance; whereas the point position along the predictor variable axis (low, medium, and high) indicates the SHapley Additive exPlanation value contribution. For each functional axis, we also show a hand-selected example of a real leaf venation network that approximates this 'optimum' combination of leaf venation architecture traits. Zoom insets for each image show small, medium, and large veins as light green, orange, and purple lines, respectively. For traits abbreviations, see Table [1.](#page-2-0)

because Sack et al. ([2008\)](#page-16-0) only damaged the midvein of palmately veined species, while we severed all midribs, canceling out any extra redundancy provided by having multiple primary veins. In both studies, treatments were an explicitly mechanical simulation of herbivory and did not include chemical signaling path-ways (Waterman et al., [2019\)](#page-17-0) that may impact  $K_{\text{leaf}}$  changes in response to herbivory.

Flow efficiency was only weakly influenced by venation architecture, suggesting that leaf hydraulics might be highly controlled by venation anatomical traits or outside-venation traits (Carin-gella et al., [2015;](#page-15-0) Xiong & Flexas, [2022](#page-17-0); Scoffoni et al., [2023\)](#page-17-0). Theoretically, venation networks with high  $MST_{small}$  (i.e. treelike minor veins) should achieve the highest water supply rate for a given investment of veins in the absence of damages (Corson, [2010](#page-15-0); Katifori et al., [2010](#page-16-0)). In our study, however, higher  $K_{\text{leaf}_{\text{max}}}$  was found in networks with intermediate MST<sub>small</sub>, perhaps because real leaves are constantly subjected to damage, and so they need to optimize not only the flow efficiency, but also resilience/resistance to damages. Importantly, flow efficiency may be more related to the architecture traits of very minor veins (i.e. veins with width < 0.01 mm), which were missing in our dataset. This could be another reason for the weak relationships between flow efficiency and venation architecture in our study.

#### Limitations and future directions

In this study, we were unable to assess veins with width  $< 10 \mu m$ . Thus, our estimation of small vein features' contribution to leaf functions may be biased. Future work should merge venation architecture statistics estimated from high-resolution images of whole leaves with those obtained from microscopic images of leaf subsections to better assess traits at different vein spatial scales. Another limitation is that our  $P_{50}$  and  $P_{88}$  values obtained using the EFM method cannot differentiate  $K_{\text{leaf}}$  declines caused by xylem embolisms from those caused by changes in the permeability of outside-xylem tissues (Sack & Scoffoni, [2012](#page-16-0)). Furthermore, due to methodological issues related to sample size (i.e.  $P_{50}$ ) and  $P_{88}$  were only measured on 32 out of 122 species), our results linking venation architecture and resistance to embolisms are provisional and need to be further explored. To better quantify the contribution of venation features to damage resistance during drought, future research should combine different methods that quantify how both inside- (Brodribb et al., [2016;](#page-15-0) Scoffoni et al., [2017a\)](#page-17-0) and outside-xylem changes (Scoffoni et al., [2023](#page-17-0)) influence the whole-leaf resistance under increasing water stress.

#### Conclusions

Variation in leaf venation architecture features across spatial scales was important to determine different leaf functions, especially mechanical support, damage resistance, and cost. However, due to the high dimensionality of the functional space, generalized leaf architecture-function trade-offs across the plant phylogeny were weak. Multiple combinations of venation and nonvenation traits were found to achieve equivalent functional performance. The absence of generalized rules linking leaf architecture-function observed here challenges prior studies that have proposed simple and generalizable models of leaf functioning. To further understand the rules linking leaf network architecture and function, future studies should focus on (1) quantifying the contribution of both inside- and outside-xylem traits on the different leaf functional axes, particularly damage resilience and flow efficiency; (2) investigating whether alternative metrics of network architecture not assessed here (e.g. vein orientation and connectivity, Mrad et al., [2021](#page-16-0)) may have stronger functional linkages; (3) constructing mechanistic models (e.g. Buckley et al., [2017](#page-15-0)) to determine the venation network architectures that are optimal under selection for different combinations of functions and costs; and (4) identifying the macroevolutionary causes of variation in network architecture across clades.

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## Competing interests

None declared.

## Authors contributions

BWB and MF acquired funding. ISM and BWB designed the study. HF provided logistical support to access the University of California Botanical Garden living collections and research glasshouse. ISM, MB, SC, AF, LG, JR and JT collected leaf samples and obtained the functional dataset. IN, AC, AE, SM, SC and NY prepared cleared leaf images for network extraction. ISM, BCJ, DK, JR, MA and SM obtained the leaf anatomical dataset. MF developed the program for leaf venation extraction, which was improved by SS and SM. ISM, BV, JM, MS and CT extracted leaf venation networks and obtained the leaf venation architectural dataset. ISM analyzed the data and drafted the manuscript. All authors contributed to and revised the manuscript.

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## Data availability

The leaf trait dataset, including both leaf venation network architecture and functional traits, as well as high-resolution cleared leaf images, and venation segmentations are permanently archived on Dryad repository at doi: [10.5061/dryad.1g1jwsv36](https://doi.org/10.5061/dryad.1g1jwsv36). Algorithms to analyze images and to calculate multiscale statistics are available in Xu et al. [\(2021](#page-17-0)). R-code to reproduce all analysis are available at [https://github.com/ilamatos/venation\\_tradeoffs/.](https://github.com/ilamatos/venation_tradeoffs/)

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## Supporting Information

Additional Supporting Information may be found online in the Supporting Information section at the end of the article.

Fig. S1 Phylogenetic tree for 122 species.

Fig. S2 Number of samples (species) with a vein present at each vein spatial scale  $(r_{\min})$ .

Fig. S3 Variation in leaf functional traits of 122 plant species across clades.

Fig. S4 Vein density variation across vein spatial scales.

Fig. S5 Minimum spanning tree ratio variation across vein spatial scales.

Fig. S6 Loop elongation ratio variation across vein spatial scales.

Fig. S7 Eigenvalues (gray bars) for each principal component with null model values generated by the broken-stick model (red line).

Fig. S8 Results of principal component analysis of leaf venation architecture and functional traits across 50 classes of vein diameter sizes.

Fig. S9 Heat maps showing the scaled importance of each predictor variable.

Fig. S10 Heat maps showing variation across models in the importance of predictor variables using scaled vein sizes.

Fig. S11 Heat maps showing variation across models in the importance of predictor variables using unscaled vein sizes.

Fig. S12 SHapley Additive exPlanation summary plots using unscaled vein sizes.

Fig. S13 SHapley Additive exPlanation summary plots using scaled vein sizes.

Fig. S14 Variable importance and SHapley Additive exPlanation values using imputed  $P_{50}$  and  $P_{88}$  trait values.

Fig. S15 Strength of interactions between predictor variables.

Methods S1 Supplementary methods.

Table S1 List of 122 plant species evaluated in this study.

Table S2 Imputed values for 14 leaf functional traits.

Table S3 Summary statistics of leaf venation architecture and functional traits.

Table S4 Principal component loadings for the leaf venation architecture and functional traits across 16 principal components.

Table S5 Principal component (PC) scores for the three first PCs for each of the 120 plant species across 50 vein width sizes  $(r_{\min})$  bins.

Table S6 Best gradient boosted models selected for each leaf functional trait.

Table S7 Performance metrics for all gradient boosted models.

Table S8 Strength (H-statistic) of pairwise interactions.

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