

A stomatal model of anatomical tradeoffs between photosynthesis and pathogen defense

Christopher D. Muir^{1*}

¹ School of Life Sciences, University of Hawaii, Honolulu, Hawaii, USA

Correspondence*:
Christopher D. Muir
cdmuir@hawaii.edu

ABSTRACT

Stomatal pores control both leaf gas exchange and are an entry for many plant pathogens, setting up the potential for tradeoffs between photosynthesis and defense. To prevent colonization and limit infection, plants close their stomata after recognizing pathogens. In addition to closing stomata, anatomical shifts to lower stomatal density and/or size may also limit pathogen colonization, but such developmental changes would permanently reduce the gas exchange capacity for the life of the leaf. I developed and analyzed a spatially explicit model of pathogen colonization on the leaf as a function of stomatal size and density, anatomical traits which determine maximum rates of gas exchange. The model predicts greater stomatal size or density increases colonization, but the effect is most pronounced when stomatal cover is low. I also derived scaling relationships between stomatal size and density that preserves a given probability of colonization. These scaling relationships set up a potential conflict between maximizing defense and minimizing stomatal cover. To my knowledge, this is the first mathematical model connecting gas exchange and pathogen defense via stomatal anatomy. It makes predictions that can be tested with experiments and may explain variation in stomatal anatomy among plants. The model is generalizable to many types of pathogens, but lacks significant biological realism that may be needed for accurate predictions.

Keywords: anatomy, leaf gas exchange, model, pathogen, photosynthesis, scaling, stomata, tradeoff

INTRODUCTION

Stomata evolved to regulate gas exchange in and out of the leaf (Hetherington and Woodward, 2003; Berry et al., 2010; Chater et al., 2017), but many plant pathogens take advantage of these chinks in the leaf cuticular armor to infect prospective hosts (Zeng et al., 2010; McLachlan et al., 2014; Melotto et al., 2017). The density and size of stomata set the anatomical maximum rate of stomatal conductance to CO₂ and water vapor (Brown and Escombe, 1900; Parlange and Waggoner, 1970; Franks and Farquhar, 2001; Franks and Beerling, 2009b; Lehmann and Or, 2015; Sack and Buckley, 2016; Harrison et al., 2019), but the pore area shrinks and expands in response to internal and external factors to regulate gas exchange dynamically (Buckley, 2019). Many plant pathogens, including viruses (Murray et al., 2016), bacteria (Melotto et al., 2006; Underwood et al., 2007), protists (Fawke et al., 2015), and fungi (Hoch et al., 1987; Zeng et al., 2010) use stomatal pores to gain entry into the leaf. Since stomatal conductance is a major limitation on photosynthesis (Farquhar and Sharkey, 1982; Jones, 1985) while pathogens reduce fitness, this sets up a potential tradeoff between increased photosynthesis and defense against pathogens. Although there have been many empirical studies on the effect that pathogens have on stomata, there is no theoretical

framework in which to place these findings. Lack of a theoretical framework makes it difficult to answer general questions about how selection for pathogen defense constrains maximum rates of gas exchange.

Stomatal anatomy is the key link between gas exchange and pathogen colonization. The density and size of stomata not only determines the theoretical maximum stomatal conductance ($g_{s,max}$), but is also proportional to the operational stomatal conductance ($g_{s,op}$) in many circumstances (Franks et al., 2009, 2014; Dow et al., 2014a; McElwain et al., 2016; Murray et al., 2019). Therefore, I use anatomical $g_{s,max}$ as a proxy for $g_{s,op}$ and do not address dynamic changes in stomatal aperture (see Discussion). Harrison et al. (2019) recently reviewed the relationship between stomatal anatomy and gas exchange in detail.

Many pathogens rely on stomata to gain entry into the leaf, but it is unclear how anatomical traits that determine $g_{s,max}$ (size and density) affect the ability of pathogens to colonize leaves. The impact of pathogens on host fitness is complex, but after a pathogen reaches a host, the first major step is colonization. Once infected, a pathogen can reduce fitness. Susceptible hosts can lose much of their biomass or die, but even resistant hosts must allocate resources to defense or reduce photosynthesis because of necrosis around sites of infection. Plants can limit colonization physiologically by closing stomata after they recognize pathogens, called stomatal defense (Melotto et al., 2017). Anatomy may be another layer of defense; plants can reduce pathogen colonization by developing leaves with lower stomatal density and/or size. Infection increases in leaves with higher stomatal density (McKown et al., 2014; Tateda et al., 2019; Dutton et al., 2019; Fetter et al., 2019), which suggests that both anatomical and physiological responses (stomatal closure) affect host colonization. By the same logic, if stomatal density were held constant, larger stomata should also increase colonization because they occupy more area. One key difference between physiological and anatomical defenses is that stomatal closure reduces gas exchange transiently, whereas anatomy constrains gas exchange throughout the life of the leaf. Because we do not understand the relationship between stomatal anatomy and colonization well, we cannot predict the relationship between $g_{s,max}$ and infection.

If stomatal size and density affect pathogen colonization, selection to limit colonization may shape stomatal size-density scaling relationships. Botanists have long recognized that stomatal size and density are inversely correlated (Weiss, 1865; Tichá, 1982; Hetherington and Woodward, 2003; Sack et al., 2003; Franks and Beerling, 2009a; Brodribb et al., 2013; de Boer et al., 2016), but the evolutionary origin of this relationship is not yet known. Here I argue that pathogens could shape selection on this relationship. Explanations for inverse size-density scaling are usually cast in terms of preserving $g_{s,max}$ and/or total epidermal area allocated to stomata (f_S ; de Boer et al., 2016) because there are many combinations of stomatal size and density that have same $g_{s,max}$ or same f_S :

$$g_{s,max} = bmDS^{0.5} \quad (1)$$

$$f_S = DS. \quad (2)$$

b and m are assumed to be biophysical and morphological constants, *sensu* Sack and Buckley (2016) (see Supplementary Material). If size and density also affect pathogen colonization, then selection from foliar pathogens could significantly alter the size-density scaling relationship. The empirical size-density scaling relationship is linear on a log-log scale, determined by an intercept α and slope β :

$$D = e^{\alpha} S^{-\beta}; \quad (3)$$

$$d = \alpha - \beta S. \quad (4)$$

69 For brevity, $d = \log(D)$ and $s = \log(S)$. Rearranging Equations 1 and 2, a scaling relationship where
70 $\beta = 0.5$ preserves $g_{s,\max}$ while $\beta = 1$ preserves f_S .

71 How would adding pathogens alter these predicted scaling relationships? For simplicity, imagine two
72 environments, one without foliar pathogens and one with lots. In the absence of foliar pathogens, we expect
73 size-density scaling to preserve $g_{s,\max}$, f_S , or some least-cost combination of them. What happens when
74 we introduce pathogens? Assuming that stomatal size and density increase pathogen colonization, then
75 selection will favor reduced size and/or density. This would change the intercept α but not the slope. The
76 effect of foliar pathogens on the slope depends on the relationship between size, density, and probability of
77 colonization. If the probability of colonization is proportional to the product of *linear* stomatal size ($S^{0.5}$)
78 and density ($\propto DS^{0.5}$ as for $g_{s,\max}$) then it has the same effect on the slope as $g_{s,\max}$ because there are many
79 combinations of D and $S^{0.5}$ that have same probability of colonization. If the probability of colonization is
80 proportional to the product of *areal* stomatal size (S) and density ($\propto DS$ as for f_S) then it has the same
81 effect on the slope as f_S because there are many combinations of D and S that have same probability of
82 colonization. Alternatively, the probability of colonization may have a different scaling relationship (neither
83 0.5 nor 1) or may be nonlinear on a log-log scale. Unlike $g_{s,\max}$ and f_S , we do not have theory to predict a
84 stomatal size-density relationship that preserves the probability of colonization.

85 In summary, the physical relationship between stomatal size, density, and conductance is well established
86 (Harrison et al., 2019). The same traits likely affect the probability of pathogen colonization, but we do not
87 have a theoretical model that makes quantitative predictions. The inverse stomatal size-density relationship
88 has usually been explained in terms of preserving stomatal conductance and/or stomatal cover, but selection
89 by pathogens might alter scaling. To address these gaps, the goals of this study are to 1) introduce a spatially
90 explicit model pathogen colonization on the leaf surface; 2) use the model to predict the relationship
91 between $g_{s,\max}$, f_S , and the probability of colonization; 3) work out what these relationships predict about
92 stomatal size-density scaling.

MODEL

93 In this section, I introduce a spatially explicit model of pathogen colonization on a leaf surface. I explain the
94 model structure and assumptions here; the Materials and Methods section below describes how I analyzed
95 the model to address the goals of the study. For generality, I refer to a generic “pathogen” that lands on leaf
96 and moves to a stomate. The model is agnostic to the type of pathogen (virus, bacterium, fungus, etc.) and
97 the specific biological details of how it moves (biotrophy). For example, motile bacterial cells can land
98 and move around where fungi may germinate from a cyst and grow until they form an appressorium for
99 infection. These very different biotrophic movements on the leaf are treated identically here. I used SymPy
100 version 1.3 (Meurer et al., 2017) for symbolic derivations.

101 **Table 1** | Glossary of mathematical symbols. The columns indicate the mathematical Symbol used in the
102 paper, the associated symbol used in R scripts, scientific Units, and a verbal Description.

Symbol	R	Units	Description
D, d	D, d	mm^{-2}	stomatal density ($d = \log D$)
f_s	f_s	none	stomatal cover ($f_s = DS$)
$g_{s,\max}$	$g_s\max$	$\text{mol m}^{-2} \text{s}^{-1}$	anatomical maximum stomatal conductance
$g_{s,\text{op}}$	$g_s\text{op}$	$\text{mol m}^{-2} \text{s}^{-1}$	operational stomatal conductance
H	H	μm^{-1}	death rate of pathogen on leaf surface
R	R	μm	stomatal radius ($S = 2\pi R^2$)
S, s	S, s	μm^2	stomatal size ($s = \log S$)
θ_i	θ_i	radians	angles between pathogen (x_p, y_p) and lines tangent to the circumference of stomate i
U	U	μm	interstomatal distance
v_i	v_i	μm	distance between pathogen (x_p, y_p) and stomate i
x_i, y_i	x_i, y_i	μm	position of stomate i
x_p, y_p	x_p, y_p	μm	starting position of pathogen

Spatial representation of stomata

Stomata develop relatively equal spacing to minimize resistance to lateral diffusion (Morison et al., 2005), allow space between stomata (Dow et al., 2014b), and prevent stomatal interference (Lehmann and Or, 2015). Here I assume that stomata are arrayed in an equilateral triangular grid with a density D and size (area) S . This assumption ignores veins, trichomes, and within-leaf variation in stomatal density. Stomata are therefore arrayed in an evenly spaced grid (Figure 1a). The interstomatal distance U , measured as the distance from the center of one stomata to the next, is the maximal diagonal of the hexagon in μm that forms an equal area boundary between neighboring stomata. The area of a hexagon is $A_{\text{hexagon}} = \frac{\sqrt{3}}{2}U^2$. By definition the stomatal density is the inverse of this area, such that $D = A_{\text{hexagon}}^{-1} = \frac{2}{\sqrt{3}}U^{-2}$. Therefore, interstomatal distance can be derived from the stomatal density as:

$$D = \frac{2}{\sqrt{3}}U^{-2}$$

$$U = \left(\frac{2}{\sqrt{3}}D^{-1} \right)^{0.5}$$

For example, if the density is $D = 10^2 \text{ mm}^{-2} = 10^{-4} \mu\text{m}^{-2}$, then U is $107.5 \mu\text{m}$. Parkhurst (1994) described this result previously. I also make the simplifying assumption that stomata are perfectly circular with radius R . This may be approximately true for fully open stomata with kidney-shaped guard cells. Although I assume stomata are circular here, in calculating $g_{s,\max}$, I assume typical allometric relationships between length, width, and pore area (see Supplementary Material).

Spatial representation of pathogen search

Now imagine that a pathogen lands at a uniform random position within the focal region and must arrive at a stomate to colonize. If it lands on a stomate, then it infects the leaf with probability 1; if it lands between stomata, then it infects the leaf with probability p_{locate} . This is the probability that it locates a stomate, which I will derive below. The probabilities of landing on or between a stomate are f_s and $1 - f_s$, respectively. Hence, the total probability of colonization is:

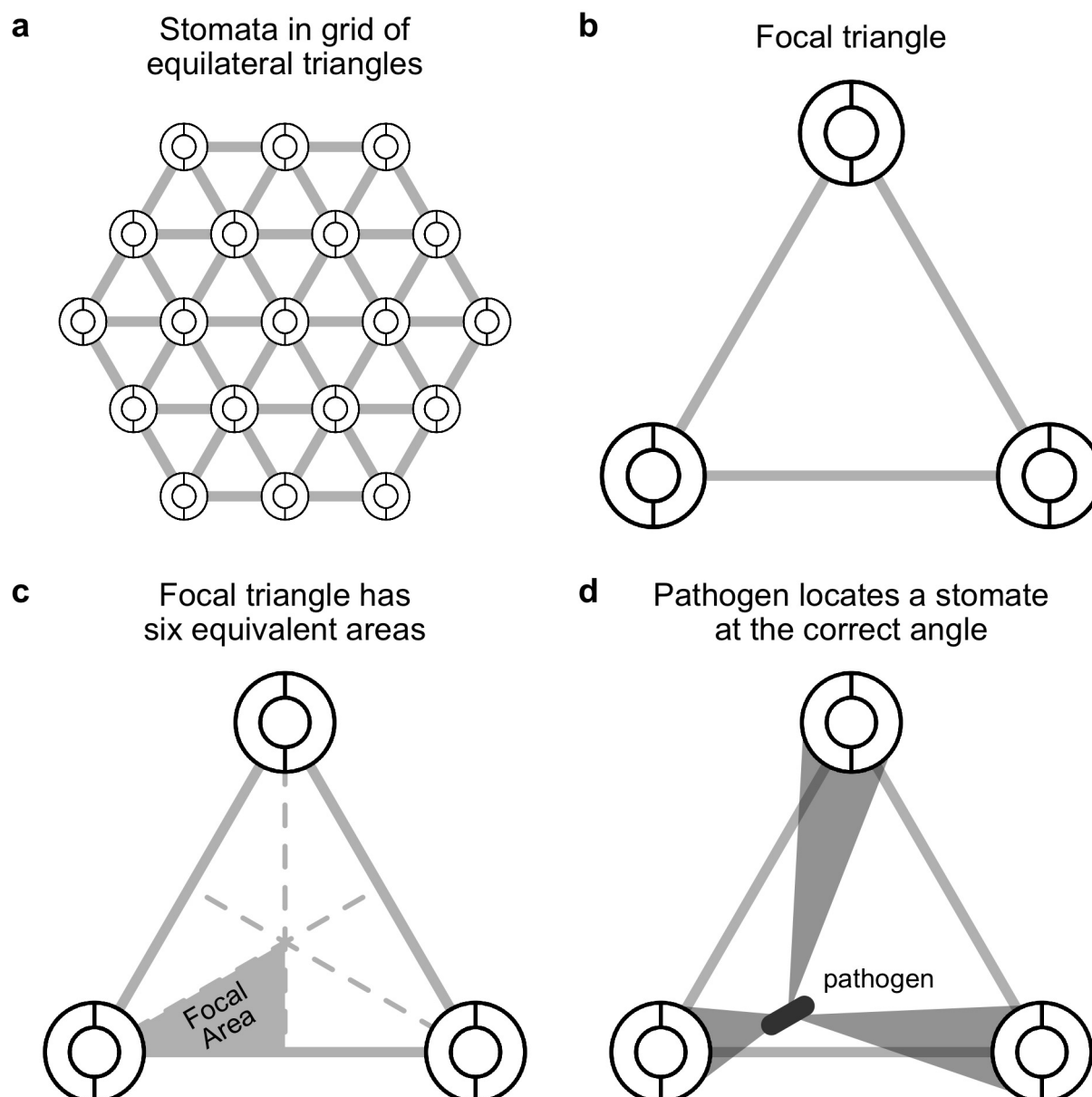


Figure 1. A spatially explicit model of stomatal anatomy and pathogen colonization. a. Stomata are assumed to be in a homogenous equilateral triangular grid, which means that we can extrapolate from b. a focal triangle to the entire leaf. The circles represent idealized stomata; the grey lines between them are for visualization. c. By symmetry, a single focal region within the focal triangle can be modeled and extrapolated to the rest of the area. d. The model assumes that a pathogen, depicted as a grey rod, lands somewhere on the leaf surface and will successfully locate a stomate if it moves at the correct angle, depicted by the grey polygons.

$$p_{\text{colonize}} = f_s + (1 - f_s)p_{\text{locate}}. \quad (5)$$

125 I assume that the pathogen cannot sense where stomata are and orients at random, thereafter traveling in
 126 that direction. If it successfully locates a stomate, it colonizes the leaf, but otherwise does not infect. If
 127 there is a high density of stomata and/or large stomata, the probability of locating a stomate increases. By

128 assuming that stomata form an equilateral triangular grid (see above), we can extrapolate what happens in
129 a focal triangle (Figure 1b) by symmetry. Further, since an equilateral triangle can be broken up into six
130 identical units (Figure 1c), we can simply calculate p_{locate} in this focal area. This implicitly assumes that
131 the probability of colonizing stomata outside the focal area is 0 because they are too far away.

132 Imagine that the pathogens lands in position (x_p, y_p) within the triangle. The centroid of the triangle
133 is at position (x_c, y_c) and a reference stomate is at position $(0, 0)$ (Figure 2a). Therefore $x_c = U/2$ and
134 $y_c = \sqrt{3}U/6$. The other stomata are at positions $(U/2, \sqrt{3}U/2)$ and $(U, 0)$ (Figure 2). x_p and y_p are defined
135 as the horizontal and vertical distances, respectively, from the pathogen to the reference stomate at position
136 $(0, 0)$.

137 Given that the pathogen starts at position (x_p, y_p) , what's the probability of contacting one of the stomata
138 at the vertices of the focal triangle? I assume the probability of contacting a stomate is equal to the
139 proportion of angular directions that lead to a stomate (Figure 1d). I solved this by finding the angles
140 $(\theta_1, \theta_2, \theta_3)$ between lines that are tangent to the outside of the three stomata and pass through (x_p, y_p)
141 (Figure 2a). If stomate i is centered at (x_i, y_i) , the two slopes of tangency as function of pathogen position
142 are:

$$t_{i,1}(x_p, y_p) = \frac{-Re_{i,2}(x_p, y_p) + e_{i,3}(x_p, y_p)}{e_{i,1}(x_p, y_p)} \quad (6)$$

$$t_{i,2}(x_p, y_p) = \frac{Re_{i,2}(x_p, y_p) + e_{i,3}(x_p, y_p)}{e_{i,1}(x_p, y_p)} \quad (7)$$

143 where

$$e_{i,1}(x_p, y_p) = (R^2 - x_i^2 + 2x_i x_p - x_p^2), \quad (8)$$

$$e_{i,2}(x_p, y_p) = \sqrt{-e_{i,1} + (y_i - y_p)^2}, \quad (9)$$

$$e_{i,3}(x_p, y_p) = -x_i y_i + x_i y_p + x_p y_i - x_p y_p. \quad (10)$$

144 Note that $i \in \{1, 2, 3\}$, indexing the three stomata in the focal triangle. The angle in radians between
145 $t_{i,1}(x_p, y_p)$ and $t_{i,2}(x_p, y_p)$ is:

$$\theta_i(x_p, y_p) = \arctan\left(\frac{t_{i,1}(x_p, y_p) - t_{i,2}(x_p, y_p)}{1 + (t_{i,1}(x_p, y_p)t_{i,2}(x_p, y_p))}\right) \quad (11)$$

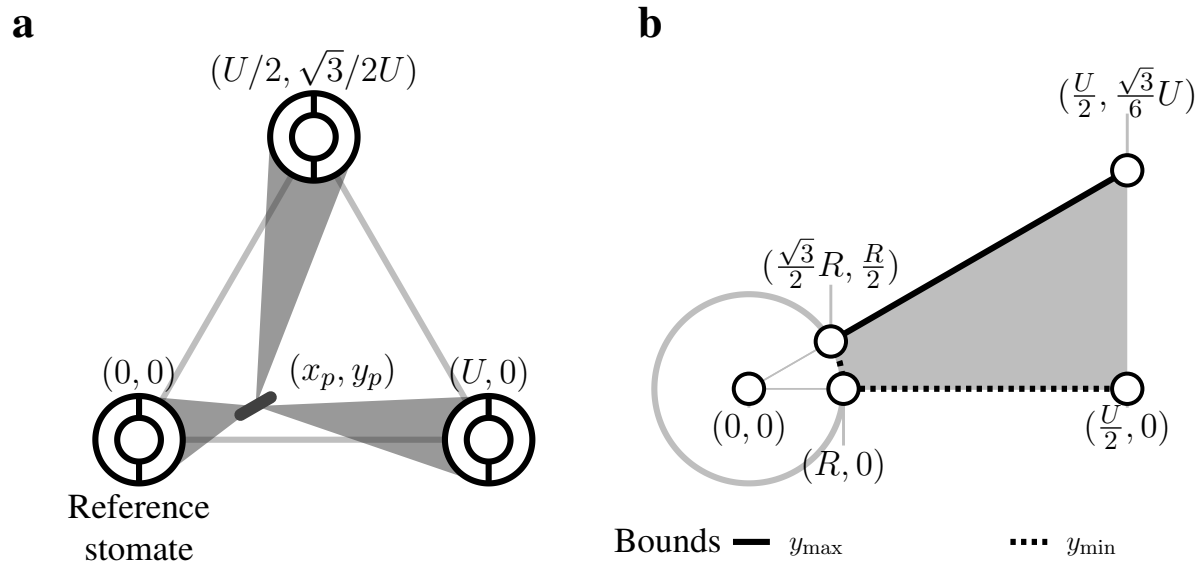


Figure 2. Spatial representation of stomata and pathogen. a. The pathogen starts at a uniform random position within the focal region denoted (x_p, y_p) . Within the focal triangle, the reference stomata is at position $(0,0)$ by definition, and other stomatal positions are determined by the interstomatal distance U . **b.** Within the focal region, a pathogen can land within the stomate (white circle with grey outline and radius R) or in the grey area. The outer borders of this area are shown and depend on R and U . For a given position x , there is a minimum y -value (y_{\min} , dashed line) and maximum y -value (y_{\max} , solid line).

I further assumed that the longer distance a pathogen must travel, the less likely it would be to locate a stomate. For example, if stomata are at very low density, then a pathogen may die before it reaches a stomate because of UV, dessication, or another factor. I included this effect by assuming the probability of reaching a stomate declines exponentially at rate H with the Euclidean distance $v_i(x_p, y_p)$ between the pathogen location and the edge of stomata i , which is distance R from its center at x_i, y_i :

$$v_i(x_p, y_p) = \sqrt{(x_i - x_p)^2 + (y_i - y_p)^2} - R. \quad (12)$$

The probability of locating a stomate as a function of x_p and y_p ($f_{\text{locate}}(x_p, y_p)$) is the sum of the angles divided by 2π , discounted by their distance from the stomate:

$$f_{\text{locate}}(x_p, y_p) = \frac{1}{2\pi} \sum_{i=1}^3 e^{-Hv_i(x_p, y_p)} \theta_i(x_p, y_p) \quad (13)$$

When $H = 0$, p_{locate} is the fraction of angles that lead from (x_p, y_p) to a stomate. When $H > 0$, p_{locate} is proportional to this fraction, but less than it depending on stomatal density, size, and starting location of the pathogen.

To obtain the average p_{locate} , we must integrate $f_{\text{locate}}(x_p, y_p)$ over all possible starting positions (x_p, y_p) within the focal area. The focal area is a 30-60-90 triangle with vertices at the center of the reference stomate $(0,0)$, the midpoint of baseline $(U/2, 0)$, and the centroid of the focal triangle $(U/2, \sqrt{3}/6U)$ (Figure 1c). Colonization occurs with probability 1 if the pathogen lands in the reference stomate, so we

160 need to integrate the probability of colonization if it lands elsewhere. This region extends from the edge of
161 the stomate, at $\sqrt{3}/2R$ to $U/2$ (Figure 2b). At any x , we integrate from the bottom of the focal area (y_{\min})
162 to the top (y_{\max}):

$$y_{\min} = f(x) = \begin{cases} \sqrt{R^2 - x^2}, & \text{if } \frac{\sqrt{3}}{2}R < x < R \\ 0, & \text{if } R \leq x \leq \frac{U}{2} \end{cases} \quad (14)$$

$$y_{\max} = f(x) = \frac{\sqrt{3}}{3}x \quad (15)$$

163 The integral is:

$$p_{\text{locate}} = \frac{1}{a_{\text{focal}}} \int_{\frac{\sqrt{3}}{2}R}^{U/2} \int_{y_{\min}}^{y_{\max}} f_{\text{locate}}(x, y) dx dy \quad (16)$$

164 a_{focal} is the area of the focal region depicted in grey in Figure 2b:

$$a_{\text{focal}} = \frac{U^2}{8\sqrt{3}} - \frac{\pi R^2}{12}$$

MATERIALS AND METHODS

165 The Model calculates a probability of host colonization (Equation 5) as a function stomatal density, size,
166 and position of a pathogen on the leaf. I solved p_{colonize} using the `integral2()` function in the **pracma**
167 package version 2.2.5 (Borchers, 2019) for numerical integration. I used R version 3.6.1 (R Core Team,
168 2019) for all analyses and wrote the paper in **rmarkdown** version 1.17 (Xie et al., 2018; Allaire et al., 2019).
169 Source code is deposited on GitHub (<https://github.com/cdmuir/stomata-tradeoff>) and
170 will be archived on Zenodo upon publication.

171 What is the relationship between stomatal size, density, and colonization?

172 I calculated p_{colonize} over a biologically plausible grid of stomatal size and density for hypostomatous
173 species based on de Boer et al. (2016). Stomatal density ranges from $10^1 - 10^{3.5} \text{ mm}^{-2}$; stomatal size
174 ranges from $10^1 - 10^{3.5} \mu\text{m}^2$. I only considered combinations of size and density where f_s was less than
175 $1/3$. For simplicity, I have not extended the current analysis to amphistomatous leaves. I crossed stomatal
176 traits with three levels of $H \in \{0, 0.01, 0.1\}$. When $H = 0$, a pathogen persists indefinitely on the leaf
177 surface. $H = 0.01$ and $H = 0.1$ correspond to low and high death rates, respectively. These values are not
178 necessarily realistic, but illustrate qualitatively how a hostile environment on the leaf surface alters model
179 predictions.

180 How do pathogens alter optimal stomatal size-density scaling?

181 The stomatal size-density scaling relationship can be explained in terms of preserving a constant $g_{s,\max}$
182 that is proportional to $DS^{0.5}$ when bm is constant (Equation 1). In other words, there are infinitely many
183 combinations of D and $S^{0.5}$ with the same $g_{s,\max}$. If $g_{s,\max}$ is held constant at C_g , then the resulting
184 size-density scaling relationship on a log-log scale is:

$$d = c_g - 0.5s$$

185 where lowercase variables are log-transformed equivalents of their uppercase counterparts. The scaling
186 exponent $\beta_g = 0.5$ preserves C_g .

187 Next, imagine there is similarly a scaling exponent β_p that preserves p_{colonize} for the product DS^{β_p} .
188 If $\beta_p = 0.5$, then p_{colonize} is always proportional to $g_{s,\text{max}}$. If $\beta_p > 0.5$, small, densely packed stomata
189 would be better defended (lower p_{colonize}) compared to larger, sparsely spaced stomata with the same
190 $g_{s,\text{max}}$. If $\beta_p < 0.5$, small, densely packed stomata would be less defended (higher p_{colonize}) compared to
191 larger, sparsely spaced stomata with the same $g_{s,\text{max}}$. I refer to the three outcomes ($\beta_p = 0.5$, $\beta_p < 0.5$,
192 and $\beta_p > 0.5$) as iso-, hypo-, and hyper-conductance, respectively. I was unable to solve analytically
193 for β_p , so I numerically calculated isoclines of p_{colonize} over the grid of D and S values described in
194 the preceding subsection. I numerically calculated the scaling relationships at a constant $p_{\text{colonize}} \in$
195 $\{0.025, 0.05, 0.1, 0.2, 0.4\}$ for $H \in \{0, 0.01, 0.1\}$.

RESULTS

196 I analyzed an idealized, spatially explicit Model of how a pathogen lands on a leaf and finds a stomate to
197 colonize the leaf using a random search. To my knowledge, this is the first model that makes quantitative
198 predictions about the relationship between stomatal anatomy, the probability of colonization, and their
199 impact on stomatal size-density scaling.

200 Nonlinear relationships between colonization, stomatal cover, and conductance

201 The probability of colonization (p_{colonize}) is not simply a one-to-one relationship between the fraction of
202 epidermal area allocated to stomata (f_s). At low f_s , p_{colonize} increases faster rapidly relative to f_s at first
203 Figure 3a). At higher f_s , the p_{colonize} increases linearly with f_s . When $H = 0$, any combination of stomatal
204 size (S) and density (D) with the same f_s have the same effect on p_{colonize} . When $H > 0$, pathogens are
205 less likely to land close enough to a stomate to infect before dying, so p_{colonize} is closer to f_s (Figure 3a).
206 Furthermore, p_{colonize} depends on D and S , not just f_s . For the same f_s , leaves with greater D have higher
207 p_{colonize} (Figure 3a). Holding f_s constant, leaves with lower D and higher S will have a greater distance
208 (v_i) between a pathogen and its stomata. When $H > 0$, this extra distance leads more pathogens to die
209 before they can find a stomate. In contrast to f_s , p_{colonize} increases at a greater than linear rate with $g_{s,\text{max}}$.
210 Greater D (smaller S) is associated with lower p_{colonize} when $g_{s,\text{max}}$ is held constant (Figure 3b). This
211 happens because p_{colonize} increases approximately linearly with S whereas $g_{s,\text{max}}$ is proportional to $S^{0.5}$.

212 Hyper-conductance size-density scaling

213 The scaling relationship between S and D that preserves p_{colonize} is always greater 0.5 (hyper-
214 conductance), but usually less than 1. When $H = 0$, the scaling relationship is essentially 1 (Figure
215 4), which means that an increase f_s leads to a proportional increase in p_{colonize} . Because the scaling
216 relationship is greater than 0.5, leaves with greater stomatal density will have lower p_{colonize} than leaves
217 lower stomatal density but the same $g_{s,\text{max}}$. In other words, increasing D and lowering S allows plants to
218 reduce p_{colonize} while maintaining $g_{s,\text{max}}$. The scaling relationship is slightly less than 1, but still greater
219 than 0.5, when $H > 0$ (Figure 4). In this area of parameter space, lower stomatal density can reduce f_s
220 while p_{colonize} is constant, but this will still result in lower $g_{s,\text{max}}$.

DISCUSSION

221 Stomatal density and size set the upper limit on gas exchange in leaves (Harrison et al., 2019) and is often
222 closely related to operational stomatal conductance in nature (Murray et al., 2019). Despite the fact that

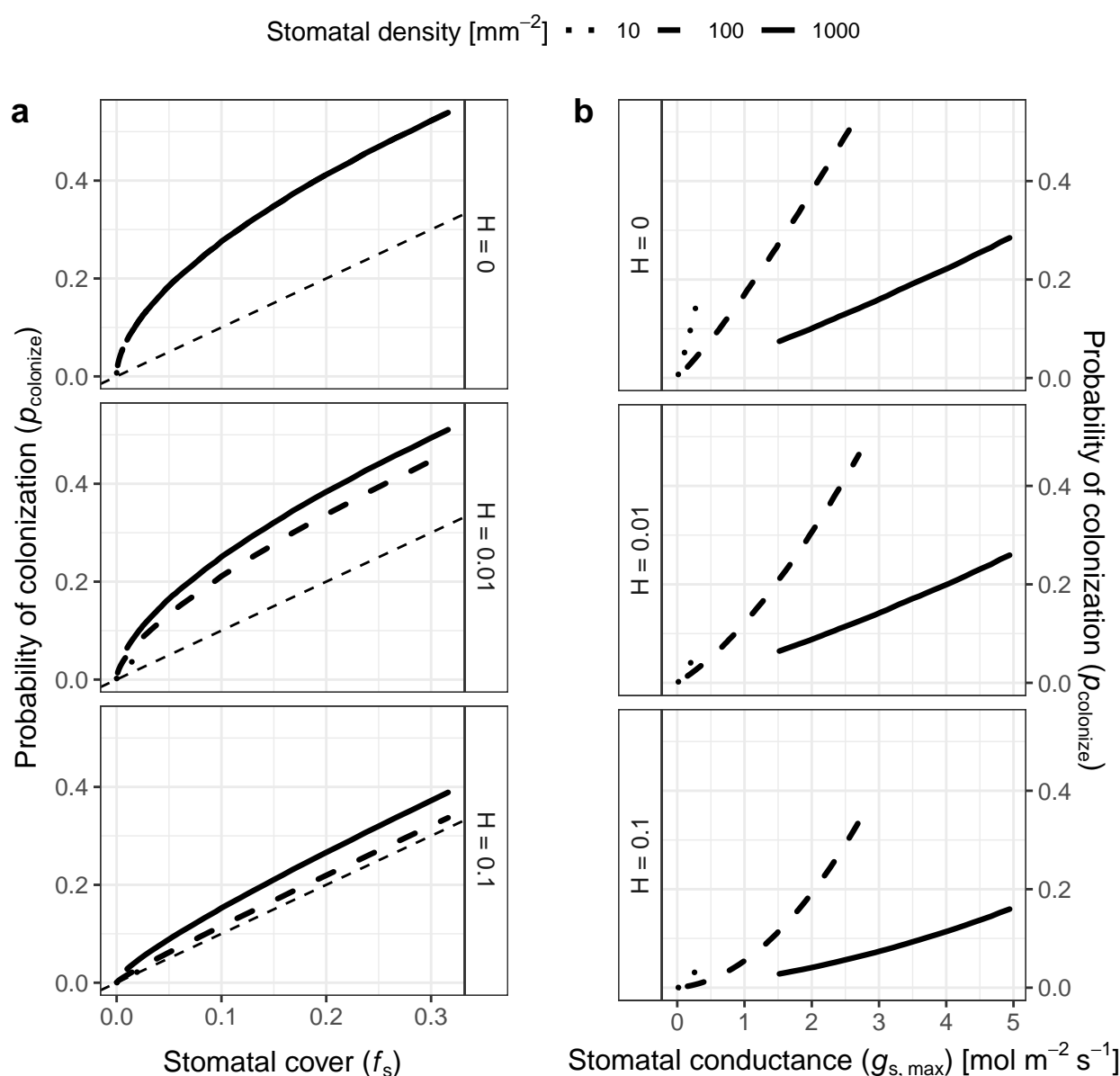


Figure 3. The probability of colonization increases with both stomatal cover and conductance a. The probability of colonization (p_{colonize} , y-axis) initially increases rapidly with stomatal cover (f_s), then slows down to a linear relationship. Overall, p_{colonize} is lower when pathogens can die on the leaf surface ($H > 0$). The relationship between f_s and p_{colonize} is the same regardless of stomatal density when $H = 0$ (upper facet). When $H > 0$, higher density (solid lines) increase p_{colonize} (lower facets). **b.** p_{colonize} increases exponentially with $g_{s, \text{max}}$ at all stomatal densities, but p_{colonize} is much lower at higher densities for a given $g_{s, \text{max}}$. The relationship between $g_{s, \text{max}}$ and p_{colonize} is similar for all values of H .

many ecologically and economically significant plant pathogens infect through stomata, the relationship between stomatal anatomy and susceptibility to foliar pathogens is less clear than it is for gas exchange. To develop testable predictions, we need mathematical models that can clarify the potential for tradeoffs between stomatal conductance, stomatal cover, and disease resistance. I used a spatially explicit model of a pathogen searching for a stomate to colonize a host. From this Model, I derived predictions about the relationship between stomatal anatomy and disease resistance for the first time. The model predicts that the

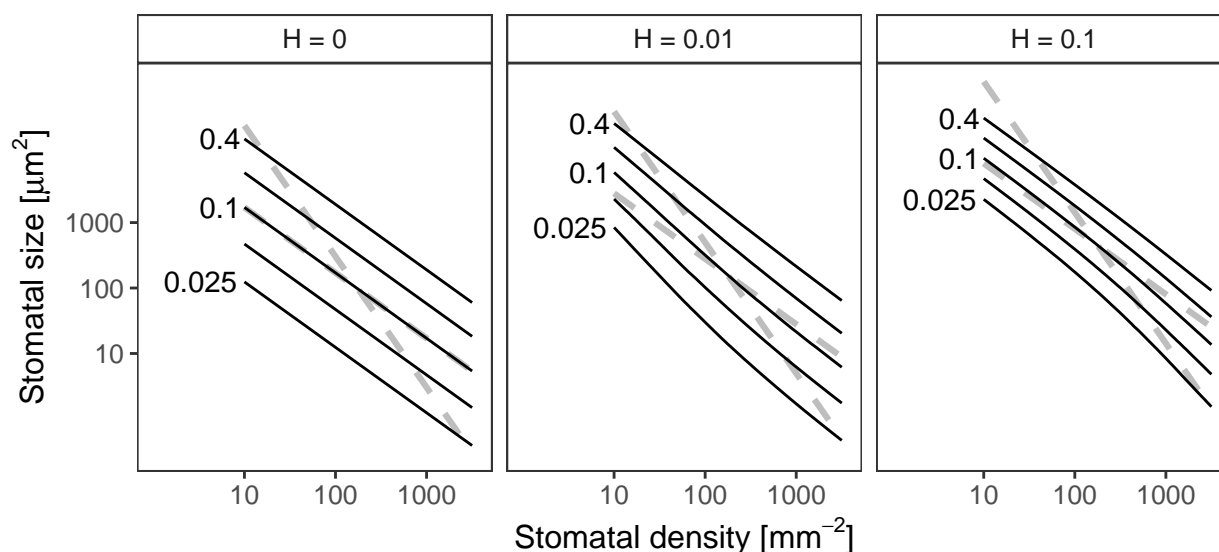


Figure 4. Log-log scaling relationships between stomatal density (D , x-axis) and size (S , y-axis) that preserve the probability of colonization (p_{colonize}). In each panel, solid lines indicate values of D and S where p_{colonize} is 0.025 (lowest line), 0.05, 0.1, 0.2, or 0.4 (highest line). For reference, dashed grey lines show scaling relationships that preserve f_s ($\beta = 1$, slope = $-1/\beta = -1$) and $g_{s,\text{max}}$ ($\beta = 0.5$, slope = $-1/\beta = -2$) drawn through the centroid of the plotting region. When the death rate on the leaf surface is low ($H = 0$), the scaling exponent is very close to $\beta = 1$. When $H > 0$, $0.5 < \beta < 1$ and is slightly nonlinear on a log-log scale.

229 probability of colonization is not always proportional to the surface area of leaf covered by stomata (f_s), as
 230 one might intuitively predict. If the leaf surface is a hostile environment and pathogens have a limited time
 231 to search, lower stomatal density decreases the probability of colonization even if f_s is constant. However,
 232 $g_{s,\text{max}}$ decreases proportionally more than the probability of colonization. The model reveals the potential
 233 for conflicting demands of maximizing disease resistance, minimizing stomatal cover, and maintaining
 234 stomatal conductance. Including the effect of anatomy on disease resistance therefore has the potential to
 235 change our understanding of how stomatal size-density scaling evolves in land plants.

236 The model predicts that in most cases, increasing stomatal cover should lead to a proportional increase
 237 in susceptibility, which is the implicit assumption of some empirical studies (e.g. McKown et al. (2014);
 238 Tateda et al. (2019); Dutton et al. (2019); Fetter et al. (2019)). It also makes new, testable predictions that
 239 are less intuitive. At very low f_s , there is a rapid increase in susceptibility (Figure 3a). If there are no
 240 stomata, the probability of colonization is 0, so the first few stomata dramatically increase the probability.
 241 This is unlikely to be significant for abaxial (lower) leaf surfaces, which usually have most of the stomata
 242 (Salisbury, 1928; Metcalfe and Chalk, 1950; Mott et al., 1984; Peat and Fitter, 1994; Jordan et al., 2014;
 243 Muir, 2015; Bucher et al., 2017; Drake et al., 2019). However, many adaxial (upper) leaf surfaces have zero
 244 or very few stomata. Using adaxial leaf surfaces, it should be possible to test if small changes in stomatal
 245 size or density have a larger effect on disease susceptibility when f_s is low. The nonlinear increase in
 246 p_{colonize} is less apparent when $H > 0$ (Figure 3a). A more hostile microenvironment (e.g. drier, higher UV)
 247 should therefore reduce the effect of increased size or density as low f_s . If true, the diminishing marginal
 248 effect of f_s on colonization could explain why stomatal ratio on the upper and lower surface is bimodal
 249 (Muir, 2015). The initial cost of adaxial (upper) stomata is high, but if the benefits outweigh the costs,
 250 then equal stomatal densities on each surface maximize CO_2 supply for photosynthesis (Parkhurst, 1978;
 251 Gutschick, 1984; Parkhurst and Mott, 1990).

An effect of stomatal size and density on susceptibility to foliar pathogens could change our understanding of stomatal size-density scaling. Since allocating leaf epidermis to stomata may be costly (Franks and Farquhar, 2007; Assmann and Zeiger, 1987; Dow et al., 2014b; Lehmann and Or, 2015; Baresch et al., 2019), selection should favor leaves that achieve a desired $g_{s,max}$ while minimizing f_S (de Boer et al., 2016). Because of their different scaling exponents (Equation 1, 2), smaller, densely packed stomata can achieve the same $g_{s,max}$ at minimum f_S . However, many leaves have larger, sparsely packed stomata. Incorporating susceptibility to disease may explain why. If pathogens have a limited time to find stomata before dying ($H > 0$), then the scaling exponent between size and density that keeps $p_{colonize}$ constant is between 0.5 and 1, the scaling exponents for $g_{s,max}$ and f_S , respectively (Figure 4). Greater density of smaller stomata can increase $g_{s,max}$ while keeping $p_{colonize}$ constant, but this will increase f_S . Conversely, f_S could decrease while keeping $p_{colonize}$ constant, but this will decrease $g_{s,max}$. This sets up the potential for conflict between competing goals. The optimal stomatal size and density will therefore depend on the precise costs and benefits of infection, stomatal conductance, and stomatal cover. This may explain why many leaves have large, sparsely packed stomata despite the fact that they could achieve the same $g_{s,max}$ and lower f_S with smaller, more densely packed stomata.

The model examines the probability of colonization for a single pathogen. The calculated probabilities of colonization should not be interpreted as exact predictions, but rather as depicting qualitative relationships between stomatal anatomy and infection severity. The model is most applicable to diseases where the host has some resistance. The energetic cost and lost photosynthetic capacity (closed stomata, necrosis, etc.) of dealing with a pathogen is assumed to be proportional to the amount of infection. The actual fitness cost will be modulated by the number of pathogens landing on the leaf and the cost of infection. In environments with fewer or less virulent pathogens, the fitness cost of infection will be less than in environments with more abundant, virulent pathogens. The model is less relevant to very susceptible host plants that can be severely damaged or killed by a small number of colonizations that spread unchecked throughout the host tissue.

The purpose of this model is to provide a general foundation to examine the relationship between stomatal size, density, and defense against foliar pathogens. In its generality, it overlooks interesting natural history and biologically important features of specific plants and their pathogens. For example, some pathogens actively seek out and find stomata (Kiefer et al., 2002), whereas I assumed that pathogens randomly orient themselves on the leaf. Including sensing should increase the probability of colonization. I also assumed that the plant does not respond by, for example, closing stomata when it senses a pathogen. However, plants can sense and close stomata when pathogens land on the leaf, but pathogens can pry stomata back open (Melotto et al., 2006). Future work on specific plant-pathogen interactions could build on this model by adding more biological realism to provide more precise predictions.

CONCLUSION

The model makes two non-intuitive predictions. First, the effect of increased stomatal density or size on susceptibility to foliar pathogens is greatest when stomatal cover is very low. Second, maximizing disease resistance sets up a potential conflict between minimizing stomatal cover and maximizing stomatal conductance. The first prediction may be relatively straightforward to test experimentally with adaxial (upper) stomata that occur at low and moderate densities within the same or closely related species (Muir et al., 2014; McKown et al., 2014; Fetter et al., 2019). The second prediction about size-density scaling is more complex because we would need to know the relationships between colonization, stomatal cover, stomatal conductance, and fitness in natural conditions. Testing these predictions in a variety of species

would help determine whether pathogens have played an important role shaping stomatal anatomy in land plants.

FUNDING

I am grateful startup funds from the University of Hawaii for supporting this work.

CONFLICT OF INTEREST STATEMENT

The author declares that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

SUPPLEMENTARY MATERIAL

I calculated $g_{s,max}$ (Equation 1) to water vapor at a reference leaf temperature ($T_{leaf} = 25^\circ \text{C}$) following Sack and Buckley (2016). They defined a biophysical and morphological constant as:

$$b = D_{wv}/v$$

$$m = \frac{\pi c^2}{j^{0.5}(4hj + \pi c)}$$

b is the diffusion coefficient of water vapor in air (D_{wv}) divided by the kinematic viscosity of dry air (v). $D_{wv} = 2.49 \times 10^{-5} \text{ m}^2 \text{ s}^{-1}$ and $v = 2.24 \times 10^{-2} \text{ m}^3 \text{ mol}^{-1}$ at 25° (Monteith and Unsworth, 2013). For kidney-shaped guard cells, $c = h = j = 0.5$.

REFERENCES

- Allaire, J., Xie, Y., McPherson, J., Luraschi, J., Ushey, K., Wickham, H., Cheng, J., Chang, W., and Iannone, R. (2019). Rmarkdown: Dynamic Documents for R. Available at: <https://github.com/rstudio/rmarkdown>.
- Assmann, S. M., and Zeiger, E. (1987). "Guard Cell Bioenergetics," in *Stomatal Function*, eds. E. Zeiger, G. D. Farquhar, and I. R. Cowan (Stanford University Press), 163–193.
- Baresch, A., Crifò, C., and Boyce, C. K. (2019). Competition for epidermal space in the evolution of leaves with high physiological rates. *New Phytologist* 221, 628–639. doi:10.1111/nph.15476.
- Berry, J. A., Beerling, D. J., and Franks, P. J. (2010). Stomata: Key players in the earth system, past and present. *Current Opinion in Plant Biology* 13, 232–239. doi:10.1016/j.pbi.2010.04.013.
- Borchers, H. W. (2019). Pracma: Practical Numerical Math Functions. R package version 2.2.5. Available at: <https://CRAN.R-project.org/package=pracma>.
- Brodribb, T. J., Jordan, G. J., and Carpenter, R. J. (2013). Unified changes in cell size permit coordinated leaf evolution. *New Phytologist* 199, 559–570. doi:10.1111/nph.12300.
- Brown, H. T., and Escombe, F. (1900). Static diffusion of gases and liquids in relation to the assimilation of carbon and translocation in plants. *Proceedings of the Royal Society of London* 67, 124–128.
- Bucher, S. F., Auerswald, K., Grün-Wenzel, C., Higgins, S. I., Garcia Jorge, J., and Römermann, C. (2017). Stomatal traits relate to habitat preferences of herbaceous species in a temperate climate. *Flora* 229, 107–115. doi:10.1016/j.flora.2017.02.011.

- 322 Buckley, T. N. (2019). How do stomata respond to water status? *New Phytologist* 224, 21–36.
323 doi:10.1111/nph.15899.
- 324 Chater, C. C. C., Caine, R. S., Fleming, A. J., and Gray, J. E. (2017). Origins and Evolution of Stomatal
325 Development. *Plant Physiology* 174, 624–638. doi:10.1104/pp.17.00183.
- 326 de Boer, H. J., Price, C. A., Wagner-Cremer, F., Dekker, S. C., Franks, P. J., and Veneklaas, E. J.
327 (2016). Optimal allocation of leaf epidermal area for gas exchange. *New Phytologist* 210, 1219–1228.
328 doi:10.1111/nph.13929.
- 329 Dow, G. J., Bergmann, D. C., and Berry, J. A. (2014a). An integrated model of stomatal development and
330 leaf physiology. *New Phytologist* 201, 1218–1226.
- 331 Dow, G. J., Berry, J. A., and Bergmann, D. C. (2014b). The physiological importance of developmental
332 mechanisms that enforce proper stomatal spacing in *Arabidopsis thaliana*. *New Phytologist* 201, 1205–1217.
333 doi:10.1111/nph.12586.
- 334 Drake, P. L., Boer, H. J., Schymanski, S. J., and Veneklaas, E. J. (2019). Two sides to every leaf: Water and
335 CO_2 transport in hypostomatous and amphistomatous
336 leaves. *New Phytologist* 222, 1179–1187. doi:10.1111/nph.15652.
- 337 Dutton, C., Hörak, H., Hepworth, C., Mitchell, A., Ton, J., Hunt, L., and Gray, J. E. (2019).
338 Bacterial infection systemically suppresses stomatal density. *Plant, Cell & Environment* 42, 2411–2421.
339 doi:10.1111/pce.13570.
- 340 Farquhar, G. D., and Sharkey, T. D. (1982). Stomatal Conductance and Photosynthesis. *Annual Review of*
341 *Plant Physiology* 33, 317–345. doi:10.1146/annurev.pp.33.060182.001533.
- 342 Fawke, S., Doumane, M., and Schornack, S. (2015). Oomycete Interactions with Plants: Infection
343 Strategies and Resistance Principles. *Microbiology and Molecular Biology Reviews* 79, 263–280.
344 doi:10.1128/MMBR.00010-15.
- 345 Fetter, K. C., Nelson, D. M., and Keller, S. R. (2019). Trade-offs and selection conflicts in hybrid poplars
346 indicate the stomatal ratio as an important trait regulating disease resistance. doi:10.1101/814046.
- 347 Franks, P. J., and Beerling, D. J. (2009a). CO_2 -forced evolution of plant gas exchange capacity and
348 water-use efficiency over the Phanerozoic. *Geobiology* 7, 227–236. doi:10.1111/j.1472-4669.2009.00193.x.
- 349 Franks, P. J., and Beerling, D. J. (2009b). Maximum leaf conductance driven by CO_2 effects on stomatal
350 size and density over geologic time. *Proceedings of the National Academy of Sciences* 106, 10343–10347.
- 351 Franks, P. J., Drake, P. L., and Beerling, D. J. (2009). Plasticity in maximum stomatal conductance
352 constrained by negative correlation between stomatal size and density: An analysis using *Eucalyptus*
353 *globulus*. *Plant, Cell & Environment* 32, 1737–1748. doi:10.1111/j.1365-3040.2009.002031.x.
- 354 Franks, P. J., and Farquhar, G. D. (2001). The effect of exogenous abscisic acid on stomatal development,
355 stomatal mechanics, and leaf gas exchange in *Tradescantia virginiana*. *Plant Physiology* 125, 935–942.
- 356 Franks, P. J., and Farquhar, G. D. (2007). The Mechanical Diversity of Stomata and Its Significance in
357 Gas-Exchange Control. *Plant Physiology* 143, 78–87. doi:10.1104/pp.106.089367.
- 358 Franks, P. J., Royer, D. L., Beerling, D. J., Van de Water, P. K., Cantrill, D. J., Barbour, M. M.,
359 and Berry, J. A. (2014). New constraints on atmospheric CO_2 concentration for the Phanerozoic:

- 360 Franks et al.: New constraints on Phanerozoic CO₂. *Geophysical Research Letters* 41, 4685–4694.
361 doi:10.1002/2014GL060457.
- 362 Gutschick, V. P. (1984). Photosynthesis model for C₃ leaves incorporating CO₂ transport, propagation of
363 radiation, and biochemistry 1. Kinetics and their parameterization. *Photosynthetica* 18, 549–568.
- 364 Harrison, E. L., Arce Cubas, L., Gray, J. E., and Hepworth, C. (2019). The influence of
365 stomatal morphology and distribution on photosynthetic gas exchange. *The Plant Journal*, tpj.14560.
366 doi:10.1111/tpj.14560.
- 367 Hetherington, A. M., and Woodward, F. I. (2003). The role of stomata in sensing and driving
368 environmental change. *Nature* 424, 901–908. doi:10.1038/nature01843.
- 369 Hoch, H. C., Staples, R. C., Whitehead, B., Comeau, J., and Wolf, E. D. (1987). Signaling for Growth
370 Orientation and Cell Differentiation by Surface Topography in Uromyces. *Science, New Series* 235,
371 1659–1662. Available at: <http://www.jstor.org/stable/1698314>.
- 372 Jones, H. G. (1985). Partitioning stomatal and non-stomatal limitations to photosynthesis. *Plant, Cell &*
373 *Environment* 8, 95–104. doi:10.1111/j.1365-3040.1985.tb01227.x.
- 374 Jordan, G. J., Carpenter, R. J., and Brodribb, T. J. (2014). Using fossil leaves as
375 evidence for open vegetation. *Palaeogeography, Palaeoclimatology, Palaeoecology* 395, 168–175.
376 doi:10.1016/j.palaeo.2013.12.035.
- 377 Kiefer, B., Riemann, M., Büche, C., Kassemeyer, H.-H., and Nick, P. (2002). The host guides
378 morphogenesis and stomatal targeting in the grapevine pathogen *Plasmopara viticola*. *Planta* 215, 387–393.
379 doi:10.1007/s00425-002-0760-2.
- 380 Lehmann, P., and Or, D. (2015). Effects of stomata clustering on leaf gas exchange. *New Phytologist* 207,
381 1015–1025. doi:10.1111/nph.13442.
- 382 McElwain, J. C., Yiotis, C., and Lawson, T. (2016). Using modern plant trait relationships between
383 observed and theoretical maximum stomatal conductance and vein density to examine patterns of plant
384 macroevolution. *New Phytologist* 209, 94–103. doi:10.1111/nph.13579.
- 385 McKown, A. D., Guy, R. D., Quamme, L., Klápště, J., La Mantia, J., Constabel, C. P., El-Kassaby,
386 Y. A., Hamelin, R. C., Zifkin, M., and Azam, M. S. (2014). Association genetics, geography and
387 ecophysiology link stomatal patterning in *Populus trichocarpa* with carbon gain and disease resistance
388 trade-offs. *Molecular Ecology* 23, 5771–5790. doi:10.1111/mec.12969.
- 389 McLachlan, D. H., Kopischke, M., and Robatzek, S. (2014). Gate control: Guard cell regulation by
390 microbial stress. *New Phytologist* 203, 1049–1063. doi:10.1111/nph.12916.
- 391 Melotto, M., Underwood, W., Koczan, J., Nomura, K., and He, S. Y. (2006). Plant Stomata Function in
392 Innate Immunity against Bacterial Invasion. *Cell* 126, 969–980. doi:10.1016/j.cell.2006.06.054.
- 393 Melotto, M., Zhang, L., Oblessuc, P. R., and He, S. Y. (2017). Stomatal Defense a Decade Later. *Plant*
394 *Physiology* 174, 561–571. doi:10.1104/pp.16.01853.
- 395 Metcalfe, C. R., and Chalk, L. (1950). *Anatomy of the dicotyledons, Vols. 1 & 2*. First. Oxford: Oxford
396 University Press.

- 397 Meurer, A., Smith, C. P., Paprocki, M., Čertík, O., Kirpichev, S. B., Rocklin, M., Kumar, A., Ivanov, S.,
398 Moore, J. K., Singh, S., et al. (2017). SymPy: Symbolic computing in Python. *PeerJ Computer Science* 3,
399 e103. doi:10.7717/peerj-cs.103.
- 400 Monteith, J. L., and Unsworth, M. H. (2013). *Principles of environmental physics: Plants, animals, and*
401 *the atmosphere*. 4th ed. Amsterdam ; Boston: Elsevier/Academic Press.
- 402 Morison, J. I. L., Emily Gallouët, Lawson, T., Cornic, G., Herbin, R., and work(s): N. R. B. R. (2005).
403 Lateral Diffusion of CO in Leaves Is Not Sufficient to Support Photosynthesis. *Plant Physiology* 139,
404 254–266. Available at: <http://www.jstor.org/stable/4281859>.
- 405 Mott, K. A., Gibson, A. C., and O’Leary, J. W. (1984). The adaptive significance of amphistomatic leaves.
406 *Plant, Cell & Environment* 5, 455–460.
- 407 Muir, C. D. (2015). Making pore choices: Repeated regime shifts in stomatal ratio. *Proceedings of the*
408 *Royal Society B: Biological Sciences* 282, 20151498. doi:10.1098/rspb.2015.1498.
- 409 Muir, C. D., Hangarter, R. P., Moyle, L. C., and Davis, P. A. (2014). Morphological and anatomical
410 determinants of mesophyll conductance in wild relatives of tomato (*solanum* sect. *Lycopersicon* , sect.
411 *Lycopersicoides* ; Solanaceae). *Plant, Cell & Environment* 37, 1415–1426. doi:10.1111/pce.12245.
- 412 Murray, M., Soh, W. K., Yiotis, C., Spicer, R. A., Lawson, T., and McElwain, J. C. (2019). Consistent
413 relationship between field-measured stomatal conductance and theoretical maximum stomatal conductance
414 in C₃ woody angiosperms in four major biomes. *International Journal of Plant Sciences*, 706260.
415 doi:10.1086/706260.
- 416 Murray, R. R., Embrow, M. S. M., Hetherington, A. M., and Foster, G. D. (2016). Plant virus infections
417 control stomatal development. *Scientific Reports* 6, 34507. doi:10.1038/srep34507.
- 418 Parkhurst, D. F. (1994). Diffusion of CO₂ and Other Gases Inside Leaves. *New Phytologist*
419 126, 449–479. Available at: <http://www.jstor.org/stable/2557929>.
- 420 Parkhurst, D. F. (1978). The Adaptive Significance of Stomatal Occurrence on One or Both Surfaces of
421 Leaves. *The Journal of Ecology* 66, 367. doi:10.2307/2259142.
- 422 Parkhurst, D. F., and Mott, K. A. (1990). Intercellular Diffusion Limits to CO₂ Uptake in Leaves: Studies
423 in Air and Helox. *Plant Physiology* 94, 1024–1032. doi:10.1104/pp.94.3.1024.
- 424 Parlange, J.-Y., and Waggoner, P. E. (1970). Stomatal Dimensions and Resistance to Diffusion. *Plant*
425 *Physiology* 46, 337–342. doi:10.1104/pp.46.2.337.
- 426 Peat, H. J., and Fitter, A. H. (1994). A comparative study of the distribution and density of stomata in the
427 British flora. *Biological Journal of the Linnean Society* 52, 377–393.
- 428 R Core Team (2019). *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R
429 Foundation for Statistical Computing Available at: <http://www.R-project.org/>.
- 430 Sack, L., and Buckley, T. N. (2016). The developmental basis of stomatal density and flux. *Plant*
431 *Physiology*, pp.00476.2016. doi:10.1104/pp.16.00476.
- 432 Sack, L., Cowan, P. D., Jaikumar, N., and Holbrook, N. M. (2003). The ‘hydrology’ of leaves: Co-
433 ordination of structure and function in temperate woody species. *Plant, Cell and Environment* 26, 1343–
434 1356. doi:10.1046/j.0016-8025.2003.01058.x.

- 435 Salisbury, E. J. (1928). On the Causes and Ecological Significance of Stomatal Frequency, with Special
436 Reference to the Woodland Flora. *Philosophical Transactions of the Royal Society B: Biological Sciences*
437 216, 1–65. doi:10.1098/rstb.1928.0001.
- 438 Tateda, C., Obara, K., Abe, Y., Sekine, R., Nekoduka, S., Hikage, T., Nishihara, M., Sekine, K.-T., and
439 Fujisaki, K. (2019). The Host Stomatal Density Determines Resistance to *Septoria gentianae* in Japanese
440 Gentian. *Molecular Plant-Microbe Interactions* 32, 428–436. doi:10.1094/MPMI-05-18-0114-R.
- 441 Tichá, I. (1982). Photosynthetic characteristics during ontogenesis of leaves 7. Stomata density and sizes.
442 *Photosynthetica* 16, 375–471.
- 443 Underwood, W., Melotto, M., and He, S. Y. (2007). Role of plant stomata in bacterial invasion. *Cellular*
444 *Microbiology* 9, 1621–1629. doi:10.1111/j.1462-5822.2007.00938.x.
- 445 Weiss, A. (1865). Untersuchungen über die Zahlen- und Grössenverhältnisse der Spaltöffnungen.
446 *Jahrbücher für Wissenschaftliche Botanik* 4, 125–196.
- 447 Xie, Y., Allaire, J. J., and Grolemond, G. (2018). *R Markdown: The definitive guide*. Boca Raton: Taylor
448 & Francis, CRC Press.
- 449 Zeng, W., Melotto, M., and He, S. Y. (2010). Plant stomata: A checkpoint of host immunity and pathogen
450 virulence. *Current Opinion in Biotechnology* 21, 599–603. doi:10.1016/j.copbio.2010.05.006.
- 451