# A stomatal model of anatomical tradeoffs between photosynthesis and pathogen defense

## Christopher D. Muir<sup>1\*</sup>

<sup>1</sup> School of Life Sciences, University of Hawaii, Honolulu, Hawaii, USA

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

### 2 ABSTRACT

1

3 Stomatal pores control both leaf gas exchange and are an entry for many plant pathogens, 4 setting up the potential for tradeoffs between photosynthesis and defense. To prevent colonization 5 and limit infection, plants close their stomata after recognizing pathogens. In addition to closing 6 stomata, anatmoical shifts to lower stomatal density and/or size may also limit pathogen 7 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 8 9 colonization on the leaf as a function of stomatal size and density, anatomical traits which 10 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 11 12 derived scaling relationships between stomatal size and density that preserves a given probability 13 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 14 15 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 16 is generalizable to many types of pathogens, but lacks significant biological realism that may be 17 needed for accurate predictions. 18

19 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; 20 21 Berry et al., 2010; Chater et al., 2017), but many plant pathogens take advantage of these chinks in the 22 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  $CO_2$ 23 and water vapor (Brown and Escombe, 1900; Parlange and Waggoner, 1970; Franks and Farquhar, 2001; 24 25 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 26 27 dynamically (Buckley, 2019). Many plant pathogens, including viruses (Murray et al., 2016), bacteria 28 (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 29 30 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 31 there have been many empirical studies on the effect that pathogens have on stomata, there is no theoretical 32

Muir et al.

#### Stomata tradeoff photosynthesis for defense

framework in which to place these findings. Lack of a theoreatical framework makes it difficult to answergeneral 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 41 42 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. 43 Once infected, a pathogen can reduce fitness. Susceptible hosts can lose much of their biomass or die, but 44 even resistant hosts must allocate resources to defense or reduce photosynthesis because of necrosis around 45 sites of infection. Plants can limit colonization physiologically by closing stomata after they recognize 46 pathogens, called stomatal defense (Melotto et al., 2017). Anatomy may be another layer of defense; 47 plants can reduce pathogen colonization by developing leaves with lower stomatal density and/or size. 48 Infection increases in leaves with higher stomatal density (McKown et al., 2014; Tateda et al., 2019; 49 Dutton et al., 2019; Fetter et al., 2019), which suggests that both anatomical and physiological responses 50 (stomatal closure) affect host colonization. By the same logic, if stomatal density were held constant, 51 larger stomata should also increase colonization because they occupy more area. One key difference 52 between physiological and anatomical defenses is that stomatal closure reduces gas exchange transiently, 53 whereas anatomy constrains gas exchange throught the life of the leaf. Because we do not understand the 54 relationship between stomatal anatomy and colonization well, we cannot predict the relationship between 55  $g_{s,max}$  and infection. 56

57 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 58 are inversely correlated (Weiss, 1865; Tichá, 1982; Hetherington and Woodward, 2003; Sack et al., 2003; 59 60 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. 61 Explanations for inverse size-density scaling are usually cast in terms of preserving  $g_{s,max}$  and/or total 62 63 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  $f_{s}$ : 64

$$g_{\rm s,max} = bmDS^{0.5} \tag{1}$$

$$f_{\mathbf{S}} = DS. \tag{2}$$

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  $\alpha$  and slope  $\beta$ :

Muir et al.

$$D = e^{\alpha} S^{-\beta}; \tag{3}$$

$$d = \alpha - \beta S. \tag{4}$$

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

How would adding pathogens alter these predicted scaling relationships? For simplicity, imagine two 71 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 we introduce pathogens? Assuming that stomatal size and density increase pathogen colonization, then 74 75 selection will favor reduced size and/or density. This would change the intercept  $\alpha$  but not the slope. The 76 effect of foliar pathogens on the slope depends on the relationship between size, density, and probability of colonization. If the probability of colonization is proportional to the product of *linear* stomatal size  $(S^{0.5})$ 77 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 78 combinations of D and  $S^{0.5}$  that have same probability of colonization. If the probability of colonization is 79 proportional to the product of *areal* stomatal size (S) and density ( $\propto DS$  as for  $f_S$ ) then it has the same 80 effect on the slope as  $f_{S}$  because there are many combinations of D and S that have same probability of 81 colonization. Alternatively, the probability of colonization may have a different scaling relationsip (neither 82 0.5 nor 1) or may be nonlinear on a log-log scale. Unlike  $q_{s,max}$  and  $f_{s,max}$  we do not have theory to predict a 83 stomatal size-density relationship that preserves the probability of colonization. 84

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

### MODEL

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

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

Muir et al.

#### Stomata tradeoff photosynthesis for defense

Symbol	R	Units	Description
D, d	D, d	$\mathrm{mm}^{-2}$	stomatal density ( $d = \log D$ )
$f_s$	f_s	none	stomatal cover ( $f_s = DS$ )
$g_{ m s,max}$	g_smax	$ m molm^{-2}s^{-1}$	$^{-1}$ anatomical maximum stomatal conductance
$g_{\mathrm{s,op}}$	g_sop	$ m molm^{-2}s^{-1}$	<sup>-</sup> bperational stomatal conductance
$H^{-}$	Н	$\mu { m m}^{-1}$	death rate of pathogen on leaf surface
R	R	$\mu$ m	stomatal radius ( $S = 2\pi R^2$ )
S, s	S, s	$\mu { m m}^2$	stomatal size ( $s = \log S$ )
$ heta_i$	theta_i	radians	angles between pathogen $(x_p, y_p)$ and lines tangent to the
			circumfrence of stomate <i>i</i>
U	U	$\mu$ m	interstomatal distance
$v_i$	v_i	$\mu$ m	distance between pathogen $(x_p, y_p)$ and stomate <i>i</i>
$x_i, y_i$	x_i,y_i	$\mu$ m	position of stomate i
$x_p, y_p$	х_р <b>,</b> у_р	$\mu$ m	starting position of pathogen

#### 103 Spatial representation of stomata

104 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, 105 2015). Here I assume that stomata are arrayed in an equilateral triangular grid with a density D and size 106 (area) S. This assumption ignores veins, trichomes, and within-leaf variation in stomatal density. Stomata 107 are therefore arrayed in an evenly spaced grid (Figure 1a). The interstomatal distance U, measured as the 108 distance from the center of one stomata to the next, is the maximal diagonal of the hexagon in  $\mu$ m that 109 forms an equal area boundary between neighbording 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, 110 111 interstomatal distance can be derived from the stomatal density as: 112

113

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

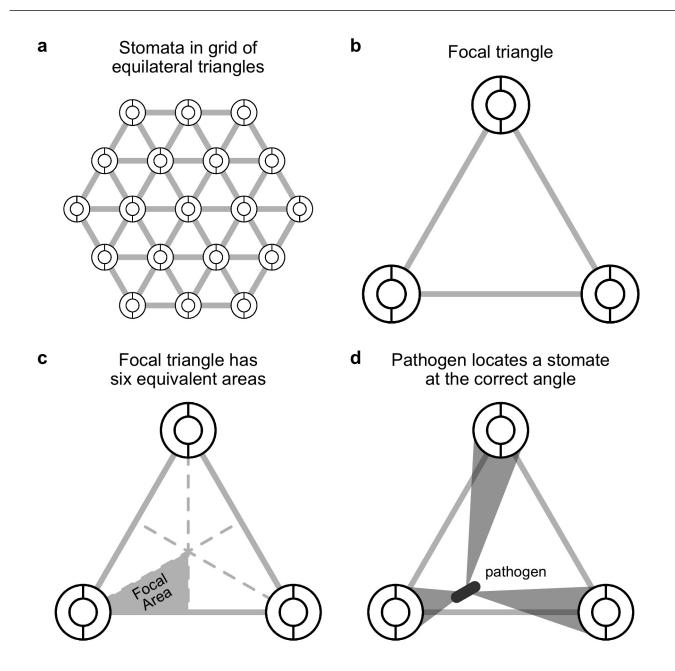
For example, if the density is  $D = 10^2 \text{ mm}^{-2} = 10^{-4} \mu \text{m}^{-2}$ , then U is 107.5  $\mu$ m. Parkhurst (1994) described this result previously. I also make the simplifying assumption that stomata are perfectly circular with radius R. This may be approximately true for fully open 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).

### 119 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  $p_{\text{locate}}$ . This is the probability that it locates a stomate, which I will derive below. The probabilities of landing on or between a stomate are  $f_{\text{S}}$  and  $1 - f_{S}$ , respectively. Hence, the total probability of colonization is:

#### Muir et al.

#### Stomata tradeoff photosynthesis for defense



**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.

$$p_{\text{colonize}} = f_{\text{S}} + (1 - f_{\text{S}}) p_{\text{locate}}.$$
(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

Muir et al.

#### Stomata tradeoff photosynthesis for defense

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

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

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

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

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

143 where

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

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

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

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

Muir et al.

#### Stomata tradeoff photosynthesis for defense

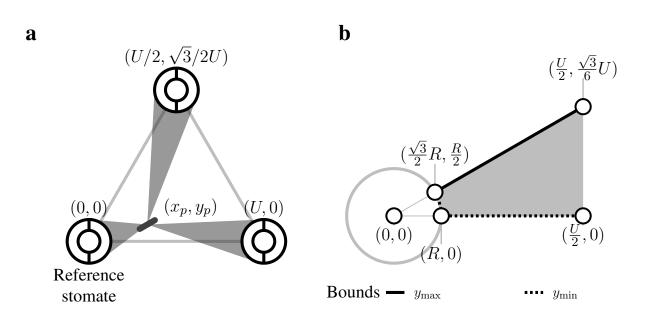


Figure 2. Spatial representation of stomata and pathogen. a. The pathogen starts at a uniform random position within the focal region denoted  $(x_p, y_p)$ . Within the focal triangle, the reference stomata is at position (0,0) by definition, and other stomatal positions are determined by the interstomatal distance U. b. Within the focal region, a pathogen can land within the stomate (white cirlce 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).

146 I further assumed that the longer distance a pathogen must travel, the less likely it would be to locate 147 a stomate. For example, if stomata are at very low density, then a pathogen may die before it reaches a 148 stomate because of UV, dessication, or another factor. I included this effect by assuming the probability of 149 reaching a stomate declines exponentially at rate H with the Euclidean distance  $v_i(x_p, y_p)$  between the 150 pathogen location and the edge of stomata i, which is distance R from its center at  $x_i, y_i$ :

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

151 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 152 divided by  $2\pi$ , discounted by their distance from the stomate:

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

153 When H = 0,  $p_{\text{locate}}$  is the fraction of angles that lead from  $(x_p, y_p)$  to a stomate. When H > 0,  $p_{\text{locate}}$  is 154 proportional to this fraction, but less than it depending on stomatal density, size, and starting location of 155 the pathogen.

To obtain the average  $p_{\text{locate}}$ , 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

#### Muir et al.

### Stomata tradeoff photosynthesis for defense

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

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

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

163 The integral is:

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

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

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

### MATERIALS AND METHODS

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

will be arcived on Zenodo upon publication. 170

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

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

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

The stomatal size-density scaling relationship can be explained in terms of preserving a constant  $g_{s,max}$ 181 that is proportional to  $DS^{0.5}$  when bm is constant (Equation 1). In other words, there are infinitely many 182 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 183 size-density scaling relationship on a log-log scale is: 184

Muir et al.

#### Stomata tradeoff photosynthesis for defense

$$d = c_a - 0.5s$$

185 where lowercase variables are log-transformed equivalents of their uppercase counterparts. The scaling 186 exponent  $\beta_q = 0.5$  preseves  $C_q$ .

187 Next, imagine there is similarly a scaling exponent  $\beta_p$  that preserves  $p_{\text{colonize}}$  for the product  $DS^{\beta_p}$ . If  $\beta_p = 0.5$ , then  $p_{\text{colonize}}$  is always proportional to  $g_{\text{s,max}}$ . If  $\beta_p > 0.5$ , small, densely packed stomata 188 189 would be better defended (lower  $p_{colonize}$ ) compared to larger, sparsely spaced stomata with the same  $g_{s,max}$ . If  $\beta_p < 0.5$ , small, densely packed stomata would be less defended (higher  $p_{colonize}$ ) compared to 190 191 larger, sparsely spaced stomata with the same  $g_{s,max}$ . 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 for  $\beta_p$ , so I numerically calculated isoclines of  $p_{\text{colonize}}$  over the grid of D and S values described in 193 the preceding subsection. I numerically calculated the scaling relationships at a constant  $p_{
m colonize} \in$ 194 195  $\{0.025, 0.05, 0.1, 0.2, 0.4\}$  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  $(p_{\text{colonize}})$  is not simply a one-to-one relationship between the fraction of epidermal area allocated to stomata ( $f_S$ ). At low  $f_S$ ,  $p_{colonize}$  increases faster rapidly relative to  $f_S$  at first 202 Figure 3a). At higher  $f_S$ , the  $p_{\text{colonize}}$  increases linearly with  $f_S$ . When H = 0, any combination of stomatal 203 size (S) and density (D) with the same  $f_{S}$  have the same effect on  $p_{\text{colonize}}$ . When H > 0, pathogens are 204 less likely to land close enough to a stomate to infect before dying, so  $p_{\text{colonize}}$  is closer to  $f_{\text{S}}$  (Figure 3a). 205 206 Furthermore,  $p_{\text{colonize}}$  depends on D and S, not just  $f_{S}$ . For the same  $f_{S}$ , leaves with greater D have higher  $p_{\text{colonize}}$  (Figure 3a). Holding  $f_{S}$  constant, leaves with lower D and higher S will have a greater distance 207  $(v_i)$  between a pathogen and its stomata. When H > 0, this extra distance leads more pathogens to die 208 before they can find a stomate. In contrast to  $f_{\rm S}$ ,  $p_{\rm colonize}$  increases at a greater than linear rate with  $g_{\rm s.max}$ . 209 Greater D (smaller S) is associated with lower  $p_{\text{colonize}}$  when  $g_{s,\text{max}}$  is held constant (Figure 3b). This 210 happens because  $p_{\text{colonize}}$  increases approximately linearly with S whereas  $g_{s,\text{max}}$  is proportional to  $S^{0.5}$ . 211

### 212 Hyper-conductance size-density scaling

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

### DISCUSSION

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

#### Stomata tradeoff photosynthesis for defense

Muir et al.

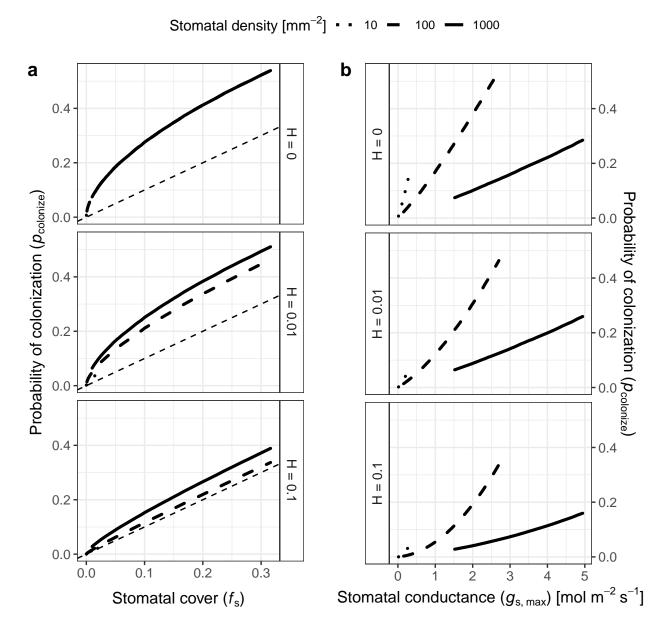


Figure 3. The probability of colonization increases with both stomatal cover and conductance a. The probability of colonization ( $p_{colonize}$ , y-axis) initially increases rapidly with stomatal cover ( $f_S$ ), then slows down to a linear relationship. Overall,  $p_{colonize}$  is lower when pathogens 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 expoentially with  $g_{s,max}$  at all stomatal densities, but  $p_{colonize}$  is much lower at higher densities for a given  $g_{s,max}$ . The relationship between  $g_{s,max}$  and  $p_{colonize}$  is similar for all values of H.

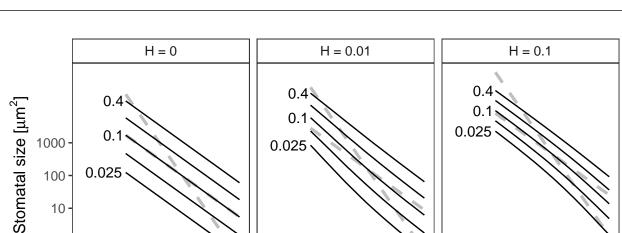
223 many ecologically and economically significant plant pathogens infect through stomata, the relationship 224 between stomatal anatomy and susceptibility to foliar pathogens is less clear than it is for gas exchange. 225 To develop testable predictions, we need mathematical models that can clarify the potential for tradeoffs 226 between stomatal conductance, stomatal cover, and disease resistance. I used a spatially explicit model of 227 a pathogen searching for a stomate to colonize a host. From this Model, I derived predictions about the 228 relationship between stomatal anatomy and disease resistance for the first time. The model predicts that the

#### Stomata tradeoff photosynthesis for defense

10

100

1000



0.025

**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  $p_{\text{colonize}}$  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  $f_{\text{S}}$  ( $\beta = 1$ , slope  $= -1/\beta = -1$ ) and  $g_{\text{s,max}}$  ( $\beta = 0.5$ , 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.

10

100

Stomatal density [mm<sup>-2</sup>]

1000

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

236 The model predicts that in most cases, increasing stomatal cover should lead to a proportional increase in susceptibility, which is the implicit assumption of some empirical studies (e.g. McKown et al. (2014); 237 Tateda et al. (2019); Dutton et al. (2019); Fetter et al. (2019)). It also makes new, testable predictions that 238 are less intuitive. At very low  $f_{\rm S}$ , there is a rapid increase in susceptibility (Figure 3a). If there are no 239 stomata, the probability of colonization is 0, so the first few stomata dramatically increase the probability. 240 This is unlikely to be significant for abaxial (lower) leaf surfaces, which usually have most of the stomata 241 (Salisbury, 1928; Metcalfe and Chalk, 1950; Mott et al., 1984; Peat and Fitter, 1994; Jordan et al., 2014; 242 243 Muir, 2015; Bucher et al., 2017; Drake et al., 2019). However, many adaxial (upper) leaf surfaces have zero or very few stomata. Using adaxial leaf surfaces, it should be possible to test if small changes in stomatal 244 size or density have a larger effect on disease susceptibility when  $f_S$  is low. The nonlinear increase in 245  $p_{\text{colonize}}$  is less apparent when H > 0 (Figure 3a). A more hostile microenvironment (e.g. drier, higher UV) 246 should therefore reduce the effect of increased size or density as low  $f_{\rm S}$ . If true, the dimishing marginal 247 effect of  $f_{\rm S}$  on colonization could explain why stomatal ratio on the upper and lower surface is bimodal 248 (Muir, 2015). The initial cost of adaxial (upper) stomata is high, but if the benefits outweigh the costs, 249 250 then equal stomatal densities on each surface maximize CO<sub>2</sub> supply for photosynthesis (Parkhurst, 1978; Gutschick, 1984; Parkhurst and Mott, 1990). 251

Muir et al.

0.025

10

100

1000

100

10

#### Muir et al.

#### Stomata tradeoff photosynthesis for defense

An effect of stomatal size and density on susceptibility to foliar pathogens could change our understanding 252 253 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., 254 2019), selection should favor leaves that achieve a desired  $g_{s,max}$  while minimizing  $f_s$  (de Boer et al., 2016). 255 Because of their different scaling exponents (Equation 1, 2), smaller, densely packed stomata can achieve 256 the same  $g_{s,max}$  at minimum  $f_s$ . However, many leaves have larger, sparsely packed stomata. Incorporating 257 susceptibility to disease may explain why. If pathogens have a limited time to find stomata before dying 258 (H > 0), then the scaling exponent between size and density that keeps  $p_{\text{colonize}}$  constant is between 0.5 259 and 1, the scaling exponents for  $g_{s,max}$  and  $f_{s}$ , respectively (Figure 4). Greater density of smaller stomata 260 can increase  $g_{s,max}$  while keeping  $p_{colonize}$  constant, but this will increase  $f_S$ . Conversely,  $f_S$  could decrease 261 while keeping  $p_{\text{colonize}}$  constant, but this will decrease  $g_{s,\text{max}}$ . This sets up the potential for conflict between 262 competing goals. The optimal stomatal size and density will therefore depend on the precise costs and 263 benefits of infection, stomatal conductance, and stomatal cover. This may explain why many leaves have 264 large, sparsely packed stomata despite the fact that they could achieve the same  $g_{s,max}$  and lower  $f_S$  with 265 smaller, more densely packed stomata. 266

The model examines the probability of colonization for a single pathogen. The calculated probabilities of 267 colonization should not be interpreted as exact predictions, but rather as depicting qualitative relationships 268 between stomatal anatomy and infection severity. The model is most applicable to diseases where the host 269 has some resistance. The energetic cost and lost photosynthetic capacity (closed stomata, necrosis, etc.) of 270 271 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 272 with fewer or less virulent pathogens, the fitness cost of infection will be less than in environments with 273 more abundant, virulent pathogens. The model is less relevant to very susceptible host plants that can be 274 severely damaged or killed by a small number of colonizations that spread unchecked throughout the host 275 tissue. 276

277 The purpose of this model is to provide a general foundation to examine the relationship between stomatal 278 size, density, and defense against foliar pathogens. In its generality, it overlooks interesting natural history 279 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 280 themselves on the leaf. Including sensing should increase the probability of colonization. I also assumed 281 282 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 283 284 (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. 285

### CONCLUSION

The model makes two non-intuitive predictions. First, the effect of increased stomatal density or size 286 on susceptibility to foliar pathogens is greatest when stomatal cover is very low. Second, maximizing 287 disease resistance sets up a potential conflict between minimizing stomatal cover and maximizing stomatal 288 conductance. The first prediction may be relatively straightforward to test experimentally with adaxial 289 (upper) stomata that occur at low and moderate densities within the same or closely related species (Muir 290 et al., 2014; McKown et al., 2014; Fetter et al., 2019). The second prediction about size-density scaling is 291 more complex because we would need to know the relationships between colonization, stomatal cover, 292 stomatal conductance, and fitness in natural conditions. Testing these predictions in a variety of species 293

#### Muir et al.

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

### FUNDING

296 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 financialrelationships 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 ( $T_{leaf} = 25^{\circ}$  C) following Sock and Buckley (2016). They defined a biophysical and morphological constant as:

$$b = D_{\rm wv}/v$$
$$m = \frac{\pi c^2}{j^{0.5}(4hj + \pi c)}$$

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 Call 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.
- 320 (2017). Stomatal traits relate to habitat preferences of herbaceous species in a temperate climate. *Flora* 321 229, 107–115. doi:10.1016/j.flora.2017.02.011.

#### Muir et al.

- Buckley, T. N. (2019). How do stomata respond to water status? *New Phytologist* 224, 21–36. doi:10.1111/nph.15899.
- Chater, C. C. C., Caine, R. S., Fleming, A. J., and Gray, J. E. (2017). Origins and Evolution of Stomatal Development. *Plant Physiology* 174, 624–638. doi:10.1104/pp.17.00183.
- de Boer, H. J., Price, C. A., Wagner-Cremer, F., Dekker, S. C., Franks, P. J., and Veneklaas, E. J.
  (2016). Optimal allocation of leaf epidermal area for gas exchange. *New Phytologist* 210, 1219–1228.
  doi:10.1111/nph.13929.
- Dow, G. J., Bergmann, D. C., and Berry, J. A. (2014a). An integrated model of stomatal development and leaf physiology. *New Phytologist* 201, 1218–1226.
- Dow, G. J., Berry, J. A., and Bergmann, D. C. (2014b). The physiological importance of developmental
  mechanisms that enforce proper stomatal spacing in *Arabidopsis thaliana*. *New Phytologist* 201, 1205–1217.
  doi:10.1111/nph.12586.
- Drake, P. L., Boer, H. J., Schymanski, S. J., and Veneklaas, E. J. (2019). Two sides to every leaf: Water and
   <span style="font-variant:Small-caps;">CO</span> 2 transport in hypostomatous and amphistomatous
   leaves. *New Phytologist* 222, 1179–1187. doi:10.1111/nph.15652.
- Dutton, C., Hõrak, H., Hepworth, C., Mitchell, A., Ton, J., Hunt, L., and Gray, J. E. (2019).
  Bacterial infection systemically suppresses stomatal density. *Plant, Cell & Environment* 42, 2411–2421.
  doi:10.1111/pce.13570.
- Farquhar, G. D., and Sharkey, T. D. (1982). Stomatal Conductance and Photosynthesis. *Annual Review of Plant Physiology* 33, 317–345. doi:10.1146/annurev.pp.33.060182.001533.
- Fawke, S., Doumane, M., and Schornack, S. (2015). Oomycete Interactions with Plants: Infection
  Strategies and Resistance Principles. *Microbiology and Molecular Biology Reviews* 79, 263–280.
  doi:10.1128/MMBR.00010-15.
- Fetter, K. C., Nelson, D. M., and Keller, S. R. (2019). Trade-offs and selection conflicts in hybrid poplars
  indicate the stomatal ratio as an important trait regulating disease resistance. doi:10.1101/814046.
- Franks, P. J., and Beerling, D. J. (2009a). CO <sub>2</sub> -forced evolution of plant gas exchange capacity and water-use efficiency over the Phanerozoic. *Geobiology* 7, 227–236. doi:10.1111/j.1472-4669.2009.00193.x.
- Franks, P. J., and Beerling, D. J. (2009b). Maximum leaf conductance driven by CO<sub>2</sub> effects on stomatal
  size and density over geologic time. *Proceedings of the National Academy of Sciences* 106, 10343–10347.
- Franks, P. J., Drake, P. L., and Beerling, D. J. (2009). Plasticity in maximum stomatal conductance
  constrained by negative correlation between stomatal size and density: An analysis using *Eucalyptus globulus*. *Plant*, *Cell & Environment* 32, 1737–1748. doi:10.1111/j.1365-3040.2009.002031.x.
- Franks, P. J., and Farquhar, G. D. (2001). The effect of exogenous abscisic acid on stomatal development, stomatal mechanics, and leaf gas exchange in *Tradescantia virginiana*. *Plant Physiology* 125, 935–942.
- Franks, P. J., and Farquhar, G. D. (2007). The Mechanical Diversity of Stomata and Its Significance in
  Gas-Exchange Control. *Plant Physiology* 143, 78–87. doi:10.1104/pp.106.089367.
- Franks, P. J., Royer, D. L., Beerling, D. J., Van de Water, P. K., Cantrill, D. J., Barbour, M. M., and Berry, J. A. (2014). New constraints on atmospheric CO <sub>2</sub> concentration for the Phanerozoic:

Muir et al.

- 360 Franks et al.: New constraints on Phanerozoic CO2. *Geophysical Research Letters* 41, 4685–4694.
  361 doi:10.1002/2014GL060457.
- Gutschick, V. P. (1984). Photosynthesis model for  $C_3$  leaves incorporating  $CO_2$  transport, propagation of radiation, and biochemistry 1. Kinetics and their parameterization. *Photosynthetica* 18, 549–568.
- Harrison, E. L., Arce Cubas, L., Gray, J. E., and Hepworth, C. (2019). The influence of stomatal morphology and distribution on photosynthetic gas exchange. *The Plant Journal*, tpj.14560. doi:10.1111/tpj.14560.
- Hetherington, A. M., and Woodward, F. I. (2003). The role of stomata in sensing and driving environmental change. *Nature* 424, 901–908. doi:10.1038/nature01843.
- Hoch, H. C., Staples, R. C., Whitehead, B., Comeau, J., and Wolf, E. D. (1987). Signaling for Growth
  Orientation and Cell Differentiation by Surface Topography in Uromyces. *Science, New Series* 235,
  1659–1662. Available at: http://www.jstor.org/stable/1698314.
- Jones, H. G. (1985). Partitioning stomatal and non-stomatal limitations to photosynthesis. *Plant, Cell & Environment* 8, 95–104. doi:10.1111/j.1365-3040.1985.tb01227.x.
- Jordan, G. J., Carpenter, R. J., and Brodribb, T. J. (2014). Using fossil leaves as evidence for open vegetation. *Palaeogeography, Palaeoclimatology, Palaeoecology* 395, 168–175. doi:10.1016/j.palaeo.2013.12.035.
- Kiefer, B., Riemann, M., Büche, C., Kassemeyer, H.-H., and Nick, P. (2002). The host guides
  morphogenesis and stomatal targeting in the grapevine pathogen Plasmopara viticola. *Planta* 215, 387–393.
  doi:10.1007/s00425-002-0760-2.
- Lehmann, P., and Or, D. (2015). Effects of stomata clustering on leaf gas exchange. *New Phytologist* 207, 1015–1025. doi:10.1111/nph.13442.
- McElwain, J. C., Yiotis, C., and Lawson, T. (2016). Using modern plant trait relationships between observed and theoretical maximum stomatal conductance and vein density to examine patterns of plant macroevolution. *New Phytologist* 209, 94–103. doi:10.1111/nph.13579.
- McKown, A. D., Guy, R. D., Quamme, L., Klápště, J., La Mantia, J., Constabel, C. P., El-Kassaby, Y. A., Hamelin, R. C., Zifkin, M., and Azam, M. S. (2014). Association genetics, geography and ecophysiology link stomatal patterning in *Populus trichocarpa* with carbon gain and disease resistance trade-offs. *Molecular Ecology* 23, 5771–5790. doi:10.1111/mec.12969.
- McLachlan, D. H., Kopischke, M., and Robatzek, S. (2014). Gate control: Guard cell regulation by microbial stress. *New Phytologist* 203, 1049–1063. doi:10.1111/nph.12916.
- Melotto, M., Underwood, W., Koczan, J., Nomura, K., and He, S. Y. (2006). Plant Stomata Function in Innate Immunity against Bacterial Invasion. *Cell* 126, 969–980. doi:10.1016/j.cell.2006.06.054.
- Melotto, M., Zhang, L., Oblessuc, P. R., and He, S. Y. (2017). Stomatal Defense a Decade Later. *Plant Physiology* 174, 561–571. doi:10.1104/pp.16.01853.
- Metcalfe, C. R., and Chalk, L. (1950). *Anatomy of the dicotyledons, Vols. 1 & 2.* First. Oxford: Oxford
  University Press.

Muir et al.

Meurer, A., Smith, C. P., Paprocki, M., Čertík, O., Kirpichev, S. B., Rocklin, M., Kumar, A., Ivanov, S.,
Moore, J. K., Singh, S., et al. (2017). SymPy: Symbolic computing in Python. *PeerJ Computer Science* 3,
e103. doi:10.7717/peerj-cs.103.

Monteith, J. L., and Unsworth, M. H. (2013). *Principles of environmental physics: Plants, animals, and the atmosphere.* 4th ed. Amsterdam ; Boston: Elsevier/Academic Press.

Morison, J. I. L., Emily Gallouët, Lawson, T., Cornic, G., Herbin, R., and work(s): N. R. B. R. (2005).
Lateral Diffusion of CO in Leaves Is Not Sufficient to Support Photosynthesis. *Plant Physiology* 139,
254–266. Available at: http://www.jstor.org/stable/4281859.

Mott, K. A., Gibson, A. C., and O'Leary, J. W. (1984). The adaptive significance of amphistomatic leaves. *Plant, Cell & Environment* 5, 455–460.

Muir, C. D. (2015). Making pore choices: Repeated regime shifts in stomatal ratio. *Proceedings of the Royal Society B: Biological Sciences* 282, 20151498. doi:10.1098/rspb.2015.1498.

Muir, C. D., Hangarter, R. P., Moyle, L. C., and Davis, P. A. (2014). Morphological and anatomical
determinants of mesophyll conductance in wild relatives of tomato (*solanum* sect. *Lycopersicon*, sect. *Lycopersicoides*; Solanaceae). *Plant, Cell & Environment* 37, 1415–1426. doi:10.1111/pce.12245.

Murray, M., Soh, W. K., Yiotis, C., Spicer, R. A., Lawson, T., and McElwain, J. C. (2019). Consistent
relationship between field-measured stomatal conductance and theoretical maximum stomatal conductance
in C<sub>3</sub> woody angiosperms in four major biomes. *International Journal of Plant Sciences*, 706260.

Murray, R. R., Emblow, M. S. M., Hetherington, A. M., and Foster, G. D. (2016). Plant virus infections
control stomatal development. *Scientific Reports* 6, 34507. doi:10.1038/srep34507.

Parkhurst, D. F. (1994). Diffusion of CO<sub>2<\sub> and Other Gases Inside Leaves. *New Phytologist*126, 449–479. Available at: http://www.jstor.org/stable/2557929.

Parkhurst, D. F. (1978). The Adaptive Significance of Stomatal Occurrence on One or Both Surfaces of
Leaves. *The Journal of Ecology* 66, 367. doi:10.2307/2259142.

Parkhurst, D. F., and Mott, K. A. (1990). Intercellular Diffusion Limits to CO<sub>2</sub> Uptake in Leaves: Studies
in Air and Helox. *Plant Physiology* 94, 1024–1032. doi:10.1104/pp.94.3.1024.

Parlange, J.-Y., and Waggoner, P. E. (1970). Stomatal Dimensions and Resistance to Diffusion. *Plant Physiology* 46, 337–342. doi:10.1104/pp.46.2.337.

Peat, H. J., and Fitter, A. H. (1994). A comparative study of the distribution and density of stomata in the
British flora. *Biological Journal of the Linnean Society* 52, 377–393.

R Core Team (2019). *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R
Foundation for Statistical Computing Available at: http://www.R-project.org/.

Sack, L., and Buckley, T. N. (2016). The developmental basis of stomatal density and flux. *Plant Physiology*, pp.00476.2016. doi:10.1104/pp.16.00476.

Sack, L., Cowan, P. D., Jaikumar, N., and Holbrook, N. M. (2003). The 'hydrology' of leaves: Coordination of structure and function in temperate woody species. *Plant, Cell and Environment* 26, 1343–
1356. doi:10.1046/j.0016-8025.2003.01058.x.

Muir et al.

Salisbury, E. J. (1928). On the Causes and Ecological Significance of Stomatal Frequency, with Special
Reference to the Woodland Flora. *Philosophical Transactions of the Royal Society B: Biological Sciences*216, 1–65. doi:10.1098/rstb.1928.0001.

Tateda, C., Obara, K., Abe, Y., Sekine, R., Nekoduka, S., Hikage, T., Nishihara, M., Sekine, K.-T., and
Fujisaki, K. (2019). The Host Stomatal Density Determines Resistance to *Septoria gentianae* in Japanese
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.
- Underwood, W., Melotto, M., and He, S. Y. (2007). Role of plant stomata in bacterial invasion. *Cellular Microbiology* 9, 1621–1629. doi:10.1111/j.1462-5822.2007.00938.x.
- Weiss, A. (1865). Untersuchungen über die Zahlen- und Grössenverhältnisse der Spaltöffnungen.
  Jahrbücher für Wissenschaftliche Botanik 4, 125–196.
- Xie, Y., Allaire, J. J., and Grolemund, G. (2018). *R Markdown: The definitive guide*. Boca Raton: Taylor
  & Francis, CRC Press.
- Zeng, W., Melotto, M., and He, S. Y. (2010). Plant stomata: A checkpoint of host immunity and pathogen
  virulence. *Current Opinion in Biotechnology* 21, 599–603. doi:10.1016/j.copbio.2010.05.006.

451