

1 Viewing stomata in action: Autonomous *in*  
2 *planta* imaging of individual stomatal movement  
3 links morphology and kinetics

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6 Summary

7 Stomata regulate plant gas exchange with the environment, balancing  
8 between water loss and CO<sub>2</sub> uptake. Gas exchange dynamics are  
9 influenced by traits such as stomatal morphology, size and density,  
10 which are commonly investigated using imprints and manual  
11 microscopy, methods that are destructive and time consuming.  
12 Moreover, these microscopic properties are statically sampled and  
13 related to the dynamic ensemble behavior: gas exchange of an entire  
14 plant or part of a leaf. Knowledge on how morphology, size and  
15 density of stomata influence the movement of individual stomata is  
16 limited. We developed a compact microscope system that can  
17 measure the kinetics of tens of stomata *in vivo* simultaneously, with  
18 sub-minute time resolution. The system can be deployed in the plant's  
19 growth environment, at minimal impact on leaf microclimate. The  
20 characteristics of our microscope and data analyses are described,  
21 and we demonstrate its capabilities on *Chrysanthemum morifolium*  
22 with novel insight into individual stomata's contribution to water-use  
23 efficiency.

24

25 Introduction

26 Due to climate-change driven rises in temperature and Vapor Pressure  
27 Deficit (VPD) (Novick *et al.*, 2016) and decreased rainfall in many  
28 agricultural areas (Dore, 2005), future crop yields are coming under  
29 pressure. Adaptation of agriculture is therefore essential to maintain  
30 food security for an increasing world population (Anderson *et al.*,

31 2020). A pillar for agricultural adaptation is the development of  
32 varieties with increased water-use efficiency (WUE). To achieve this,  
33 we need to thoroughly understand the role that stomata play in plant  
34 WUE, defined as the rate of photosynthesis divided by the rate of  
35 transpiration (Bertolino *et al.*, 2019; Buckley, 2019; Lawson & Vialet-  
36 Chabrand, 2019; Nadal & Flexas, 2019), to breed for stomatal traits  
37 underlying high WUE.

38 Stomata are the microscopic pores on plant leaves that  
39 regulate their gas exchange, by dynamically opening and closing in  
40 response to environmental and intrinsic stimuli (Lawson & Matthews,  
41 2020). Their dynamic behavior (Lawson & Vialet-Chabrand, 2019),  
42 morphology and density (Bertolino *et al.*, 2019; Duursma *et al.*, 2019)  
43 are important for plant WUE. Stomatal conductance to CO<sub>2</sub> diffusion  
44 into the leaf facilitates photosynthetic CO<sub>2</sub> fixation in the mesophyll,  
45 and stomatal movement broadly aims to balance the CO<sub>2</sub> taken up by  
46 photosynthesis with the water vapor lost through transpiration. Light  
47 intensity changes of ~25-50 fold on single leaves are frequent in the  
48 crop growth environment (Kaiser *et al.*, 2018). When the light  
49 intensity drops, the photosynthetic demand for CO<sub>2</sub> decreases near-  
50 instantaneously, while stomatal closure proceeds much more slowly.  
51 Hence, stomata that at that moment are open more than strictly  
52 necessary for CO<sub>2</sub> demand, waste water through unnecessary  
53 transpiration. Likely, fast-closing stomata are thus more water use  
54 efficient.

55 Stomatal clustering and density have major effects on WUE. The  
56 tendency to form clusters of stomata, compared to evenly spaced out  
57 stomata, negatively affects their function (and thereby WUE), by  
58 decreasing their effective response speed (Lehmann & Or, 2015).  
59 Higher stomatal density negatively impacts WUE, by increasing  
60 stomatal conductance under dark conditions, because stomata are  
61 often not fully closed (Duursma *et al.*, 2019). However, high stomatal  
62 density may increase stomatal speed and thereby WUE in fluctuating  
63 light, as it tends to correlate with small stomatal size (Bertolino *et al.*,

64 2019; Lawson & Matthews, 2020), indicating the complexity of  
65 stomatal trait effects on WUE.

66 An increase in light intensity quickly causes photosynthetic demand  
67 for CO<sub>2</sub> to increase, as biochemical limitations are lifted in the first  
68 minutes of photosynthetic induction (Sakoda *et al.*, 2021). To meet  
69 this demand for CO<sub>2</sub>, stomata need to open to increase the rate of  
70 diffusion of CO<sub>2</sub>, leading to a transient limitation of photosynthesis  
71 while opening. Transient limitations of photosynthesis in shade-sun  
72 transitions cost an estimated 10-40% of potential crop CO<sub>2</sub>  
73 assimilation (Long *et al.*, 2022). More efficient, fast-responding  
74 stomata could thus result in a CO<sub>2</sub> assimilation increase (McAusland  
75 *et al.*, 2016; Xiong *et al.*, 2018, 2022; Deans *et al.*, 2019; Acevedo-  
76 Siaca *et al.*, 2020, 2021; De Souza *et al.*, 2020; Taylor *et al.*, 2020;  
77 Eyland *et al.*, 2021), leading to similar potential increases in crop yield  
78 (Garcia *et al.*, 2023).

79 Stomatal dynamics are most frequently studied using leaf-level gas  
80 exchange measurements, and are then related to static microscopic  
81 observations of e.g. leaf epidermal peels to stomatal density and  
82 morphology. Measurements of bulk stomatal behavior, such as those  
83 of leaf gas exchange, hide the variation in dynamics between  
84 individual stomata (Kaiser & Paoletti, 2014), and studies that have  
85 resolved individual stomatal dynamics are relatively rare (Kaiser &  
86 Kappen, 1997, 2000, 2001; Kaiser, 2009; Grantz *et al.*, 2018).  
87 Moreover, minimizing the boundary layer of still air surrounding the  
88 leaf, as is common in gas exchange measurements (Busch *et al.*, 2024),  
89 limits the study of stomatal behavior under frequently occurring  
90 natural conditions, when a significant boundary layer is present.  
91 Therefore, information is lacking on how local morphology and  
92 density influence the ensemble of individual stomatal dynamics. Such  
93 characteristics could lead to a better understanding of stomatal  
94 control and thereby to breeding targets for WUE and yield (Haworth  
95 *et al.*, 2021). Additionally, it is essential to study such characteristics  
96 in the field, e.g. to understand how plants with altered stomatal

97 characteristics respond to multiple stresses in different  
98 developmental phases (Bertolino *et al.*, 2019).

99 Here, we describe a newly developed portable microscope, which can  
100 measure the opening and closure of tens of individual stomata  
101 simultaneously in the growth environment. Our method innovates on  
102 previous methods by selective use of green light for imaging and blue  
103 and red light as actinic light. In addition, we imaged the entire field of  
104 view (FOV) of the leaf's surface. We did this by creating large image  
105 stacks that were used to create leaf surface projections, in contrast to  
106 autofocus of single pores. This enabled us to relate stomatal dynamics  
107 to microscopic characteristics of the stomatal environment on the  
108 epidermis such as the distance to other stomata. Acquisition and  
109 analyses were largely automated, facilitating easy measurements. We  
110 demonstrated our method on leaves of *Chrysanthemum morifolium*,  
111 a greenhouse crop whose stomatal behavior can limit vase life  
112 (Fanourakis *et al.*, 2021) and the rate of photosynthetic induction  
113 (Zhang *et al.*, 2022).

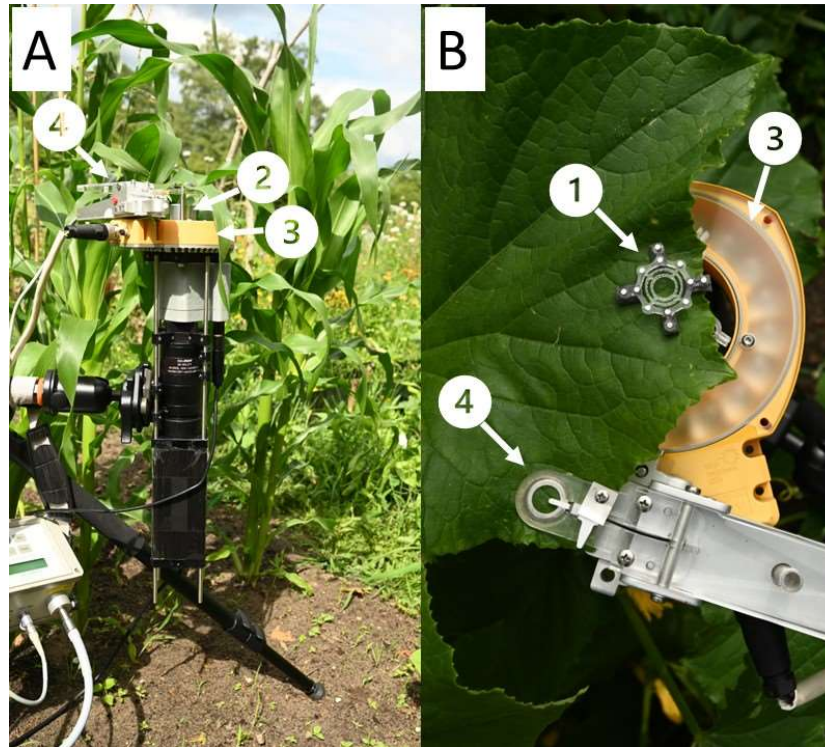
## 114 Materials and Methods

### 115 Microscope

116 The design of our microscope focused on obtaining images with sub-  
117 micrometer resolution at different focus planes to resolve the 3D  
118 surface of the leaf with sub-minute time resolution. Additionally, the  
119 design aimed to be minimally invasive to the leaf's microclimate.

120 We achieved high optical resolution with high quality optics  
121 (plan APO series objectives, Mitutoyo, Japan and TTL200-S8 tube lens,  
122 Thorlabs, United States), verified with a calibration target (micro V2,  
123 Opto, Germany). The 20x objective (NA 0.42) enabled imaging of  
124  $\sim 0.45 \text{ mm}^2$  ( $722 \times 625 \mu\text{m}$ ) of the leaf's surface. High time resolution of  
125 stacks of 100 images spaced  $1 \mu\text{m}$  apart on the axis parallel to the  
126 imaging plane (z-axis; below the  $1.6 \mu\text{m}$  depth of focus of the  
127 microscope objective) was achieved with a fast and accurate step  
128 motor (Z812, Thorlabs, United States, range of 12 mm,  $0.2 \mu\text{m}$

129 precision), which was controlled with autofocus software and a CMOS  
130 camera (BFS-U3-244S8M-C; Teledyne-FLIR, Wilsonville, OR, USA). The  
131 high sensitivity of the camera and on-camera pixel binning (2x2)  
132 allowed us to achieve good quality images at a relatively low light  
133 intensity ( $50 \pm 10 \mu\text{mol photons m}^{-2} \text{s}^{-1}$ ) and integration time ( $400 \pm$   
134  $100 \text{ ms}$ ). The use of long working distance objectives (20 mm) and leaf  
135 clips that were laser-cut from transparent polycarbonate with 2 mm  
136 neoprene cushions allowed us to clamp and image the leaf with low  
137 obstruction to air flow and light. Imaging was strictly done with green  
138 light (Effiring 525 nm; Effilux, Hürth-Efferen, Germany) in darkfield by  
139 filtering light that passed through the objective with a bandpass filter  
140 (FB550-40,  $550 \pm 8 \text{ nm}$ , FWHM  $40 \pm 8 \text{ nm}$ ; Thorlabs, Newton, NJ, USA).  
141 Monochromatic light was chosen, because it boosted image quality  
142 by limiting chromatic aberrations. Green light was selected because  
143 of its high reflectance and transmission by the leaf, higher optical  
144 resolution achievable compared to near-infrared and a maximum  
145 sensitivity of the camera in the green waveband. Optimization of  
146 imaging light was achieved before each automated measurement by  
147 manually changing the emission angle and diffusivity (by changing the  
148 opacity of the window) of the imaging light. Blue-red actinic  
149 illumination was emitted by a LED lamp (444/661 nm, 20/12 nm  
150 FWHM, 52/48%; Seven steps to heaven, the Netherlands). Both  
151 actinic and imaging light were controlled via LabVIEW (2018, National  
152 Instruments, Austin, TX, USA) and intensity calibrated at the leaf  
153 position with a PAR quantum sensor (Li-190R; Li-Cor Biosciences,  
154 Lincoln, NE, USA). Leaves were clamped in the microscope leaf holder  
155 equipped with embedded magnets at four contact points. The  
156 portable microscope was mounted on a tripod and therefore  
157 adjustable in height and angle to target leaves in a plant canopy (Fig.  
158 1). Leaf Temperature and PAR intensity fluctuations during the  
159 measurement were recorded with a leaf clip holder (2020-B)  
160 connected to a portable fluorometer (MIN-PAM; Walz, Effeltrich,  
161 Germany).



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Figure 1. Example of the microscope in use at the vegetable garden of the University of Twente (the Netherlands). A. Side view of the entire microscope with a leaf of *Zea mays* clamped in the holder. B. Top view of a leaf of *Cucumis sativa* clamped in the holder. The leaf clip (1) holds the leaf in place via magnets embedded in the polycarbonate framework. Neoprene cushions ensure a minimal effect of the clamping on the surface of the leaf. The microscope objective (2) can focus automatically and the ring light (3) provides illumination for microscopy. The leaf clip from the mini-PAM (4) records fluctuations in PAR as well as ambient air temperature (A) or leaf temperature (B).

## 173 Software

### 174 Acquisition

175 The data acquisition software was developed in Labview. Camera  
176 settings, number of images per stack with different focus planes on  
177 the z-axis (z-stack) and stack depth, imaging and stimulus light  
178 intensity and duration, as well as autofocus settings that determined  
179 the position of the stack's center (methods: Roberts, Sobel, Gradient  
180 (Lthi *et al.*, 2010)) were controlled via Labview's user interface. Full  
181 protocol files with imaging, light timing as well as light intensity  
182 settings can be loaded into the software with a graphical user  
183 interface for inspection.

### 184 Analyses

185 Acquired z-stack images, in 16bit TIFF format belonging to each z-  
186 stack, were processed in several stages (Fig. 2) using Fiji open-source



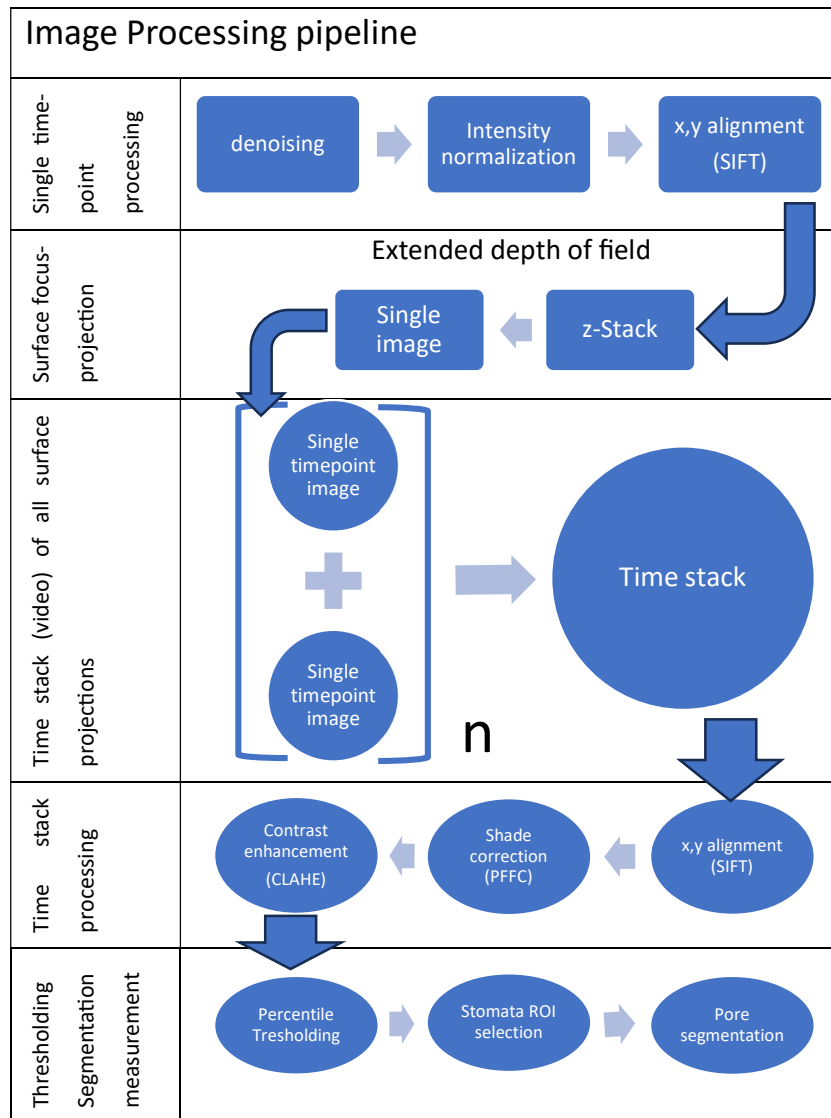
187 software (Schindelin et al., 2012). Images were denoised (despeckle  
188 and outlier removal), normalized for intensity (enhance contrast,  
189 normalize) and aligned (SIFT linear stack alignment (Lowe, 2004)). SM  
190 movie 1 is an example of a processed image stack. Next, the plugin  
191 “extended depth of field” (Forster *et al.*, 2004) was used to generate  
192 a focus projection of the leaf surface, reducing the stack to a single  
193 image per timepoint. All timepoint focus projections were then  
194 stacked again to generate a video of the entire experiment that was  
195 again aligned to adjust for x-y axis movements of the leaf during the  
196 experiment (SIFT linear stack alignment). This stack was then cropped,  
197 to include only the area that was within view during the entire  
198 experiment, shadow corrected with the pseudo flat field correction in  
199 the BioVoxel plugin (Brocher, 2015) and contrast enhanced using the  
200 CLAHE algorithm (Reza, 2004) (Fig. 3 A-C). SM movie 2 is an example  
201 of a processed video from an entire experiment, while SM movie 3 is  
202 a zoomed version of a single stoma. Auto thresholding was done in  
203 Percentile mode (Doyle, 1962), as it generated the best segmentation  
204 of open pores for *Chrysanthemum morifolium* (Fig. 3D). SM movie 4  
205 is the same video as SM movie 3 after thresholding. Stomata in the  
206 stack were manually selected with the elliptical selection tool and  
207 added in the region of interest (ROI) manager (ImageJ). Pore areas  
208 were then quantified via the ‘analyse particles’ menu for each stoma  
209 in the ROI manager, to quantify pore area and to generate masks for  
210 visual inspection of pore shapes for the entire experiment (Fig 3D).  
211 SM video 5 shows the masks generated from the stoma in SM videos  
212 3 and 4.

213 Kinetics of stomatal aperture changes were loaded into Origin  
214 (OriginLab Corporation, Northampton, MA, USA) and, data of opening  
215 and closing were fitted separately with the model (Violet-Chabrand *et*  
216 *al.*, 2013):

217 *Equation 1*

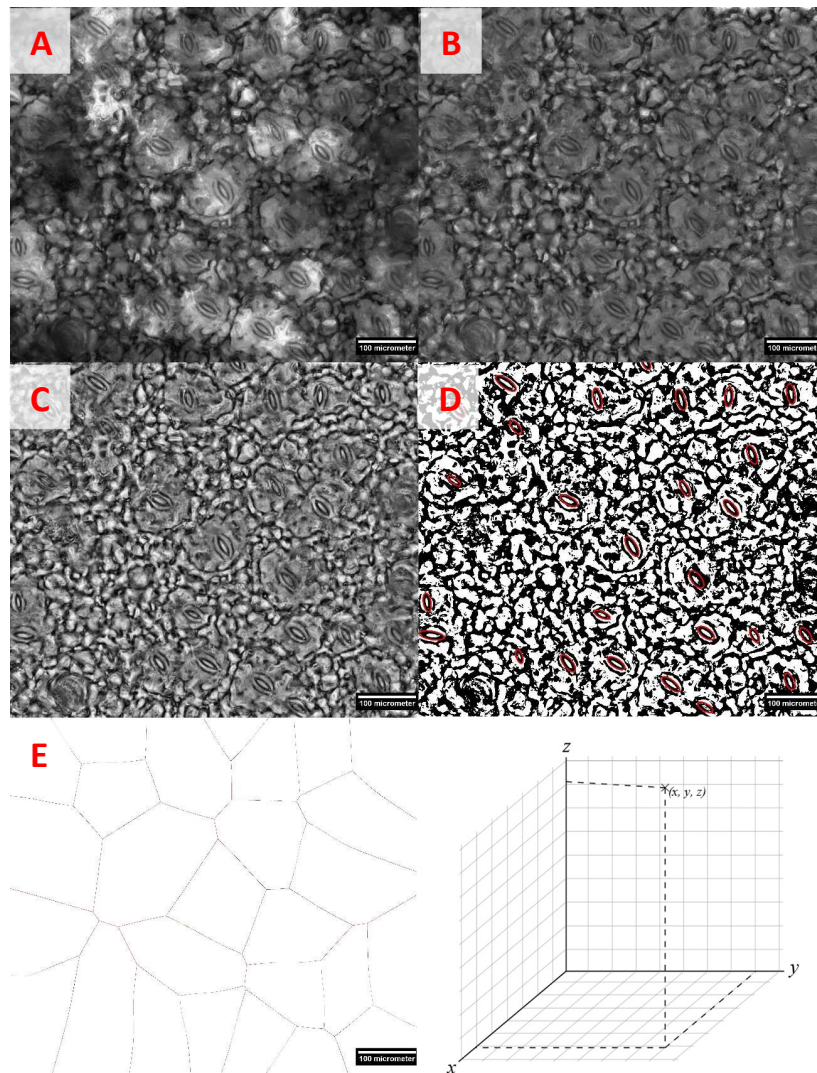
$$218 \quad A(t) = (A_{max-i} - A_{0-i}) e^{-e^{\left(\frac{\lambda_i - t}{k_i} - 1\right)}} + A_{0-i}.$$

219



220 *Figure 2. Flow chart of image processing pipeline. Single images belonging to an*  
 221 *image stack were denoised and their contrast was enhanced by intensity*  
 222 *normalization. Images with different z-axis position in a stack were then aligned to*  
 223 *correct for minor x, y movements (surface parallel to the microscope's objective). The*  
 224 *extended depth of field plugin processed the stack to create a single focus projection*  
 225 *of leaf surface. Each surface projection timepoint was added together in a video of*  
 226 *the entire experiment. The images in these videos were again aligned to correct for x,*  
 227 *y movements, shade corrected and their contrast was enhanced. Automated*  
 228 *thresholding of time stacks was done with Percentile mode. Stomatal positions were*  
 229 *manually selected with the elliptical selection tool, pore area was segmented, and*  
 230 *quantified with the analyze particles menu.*





231

232 *Figure 3. Example of time-stack processing with a single surface projection. A. Surface*  
233 *projection output from extended depth of field. B. Image in A processed with pseudo*  
234 *flat field correction (PFFC). C. Image in B processed with contrast limited adaptive*  
235 *histogram equalization (CLAHE). D. Image in C after thresholding with the Percentile*  
236 *method. Red ROIs are drawn manually with the elliptical selection tool. E Voronoi*  
237 *diagram drawn based on the elliptical selections of stomata in D. The scale bar*  
238 *indicates 100 micrometer.*

239 With  $A(t)$  the stomatal aperture at time  $t$ ,  $A_{\max}$  the maximum aperture  
240 at steady-state,  $A_i$  the initial aperture at steady-state,  $\lambda$  the time lag of  
241 the response (min) and  $k$  the time constant (min), a measure of the  
242 rapidity of the response. For stomatal closure,  $A_{\min-d}$  was defined as  
243 the final aperture at steady-state with  $A_{0-d}$ , the aperture at the start  
244 of the closing response.  $k$  and  $\lambda$  were separately quantified for open  
245 and closing responses, using subscripts  $i$  and  $d$ , respectively.

246 Equation 2

$$247 \quad A(t) = (A_{min-d} - A_{0-d}) e^{-e^{\left(\frac{\lambda_d - t}{k_d} - 1\right)}} + A_{0-d}.$$

248 Quality of fit was assessed by reduced chi squared statistic of the fit  
249 (ideally close to 1), the structure in the residuals (ideally random noise  
250 without structure) and error estimates of the fitted parameters  
251 (ideally <10%, but no larger than 100%). SM Fig. 1 provides an  
252 example of fit of the kinetics for a single stoma.

253 Some stomata were excluded from data analysis. Reasons to exclude  
254 stomata were: a) Bad segmentation due to low local image quality e.g.  
255 stoma shaded by trichome, b) incomplete stomatal pore on the edge  
256 of an image, c) lack of pore opening after a light intensity change, d)  
257 low-quality fit, as judged by  $\chi^2$ , residuals and uncertainty (>100% of  
258 value) in the fitted parameters.

259 Guard cell length (GCL) was measured in Fiji by manual use of the  
260 straight line tool in the image at t=90 min, the timepoint where  
261 stomatal aperture was generally maximal.

262 A Voronoi plot (Fig. 3E), which connects lines with equal distance to  
263 the borders of each neighbouring stoma, was created via the Voronoi  
264 tool (in the Fiji software), and each surface was measured via the  
265 'analyse particles' menu. Resulting data were used to test for  
266 relationships between the leaf area that could be assigned to a given  
267 stoma ('Voronoi area') and that stoma's aperture and kinetics. Only  
268 Voronoi surface areas that were fully inside the image were  
269 considered in this analysis.

270

271 Because our method depends on the automated focus projection of  
272 the z-stack (the reduction into a single image that represents the  
273 entire surface within the FOV), we compared it to the human  
274 operator: manual selection of the best focus position per pore. We  
275 compared the results for ten stomata with a good distribution within  
276 the FOV for 100 stacked images per 119 time points during the long  
277 term shade-sun-shade transition. Further image processing,

278 segmentation and quantification of the pore area were automated  
279 and following the identical protocol for both manual selection and  
280 focus projection.

## 281 Plant material and growing conditions

### 282 *Chrysanthemum*

283 Experiments were conducted in a growth chamber equipped with  
284 nine dimmable LED modules (DRWFR\_RSE 400V 1.1D MP; Signify,  
285 Eindhoven, the Netherlands) that produced a diurnal average of ~250  
286  $\mu\text{mol m}^{-2} \text{s}^{-1}$  of photosynthetically active radiation (PAR) at plant  
287 height. A sinusoidal light pattern (16h light period, minimum PAR of  
288  $120 \mu\text{mol m}^{-2} \text{s}^{-1}$ , maximum PAR of  $1200 \mu\text{mol m}^{-2} \text{s}^{-1}$ ) was applied with  
289 random drops in light intensity that mimicked natural irradiance  
290 fluctuations. The timing and extent of drops in light intensity changed  
291 daily, but the sinusoidal pattern and daily light sum remained fixed.  
292 Temperatures were set at 23/20 °C (day/night). Relative air humidity  
293 was 70%. Ventilation was  $0.28\text{-}0.55 \text{ m s}^{-1}$  of laminar flow.  
294 *Chrysanthemum morifolium* (cv. Anastasia; Deliflor Chrysanten,  
295 Maasdijk, the Netherlands) plants that had been grown in plastic pots  
296 (diameter 14 cm, filled in with potting soil) were cut back at the third  
297 or fourth node to allow for the formation of new axillary buds. Plants  
298 were irrigated twice per day (at 7:00 and 19:00 h) with nutrient  
299 solution, using an automatic ebb and flow system.

### 300 Stomatal aperture measurements

301 Stomatal apertures were measured on fully expanded leaves in the  
302 middle between the leaf edge and the midrib. Care was taken not to  
303 include veins in the FOV of the microscope, in order to maximize the  
304 number of stomata in the FOV. A screen was used to shade the  
305 measured plant from direct growth chamber lighting during aperture  
306 measurements. Leaves were acclimated to  $50 \mu\text{mol m}^{-2} \text{s}^{-1}$  blue-red  
307 light and  $50 \mu\text{mol m}^{-2} \text{s}^{-1}$  green imaging light for 60 minutes before  
308 image acquisition was started. Image stacks were acquired every  
309 minute for 120 minutes. After 30 minutes, the light intensity was  
310 increased to  $1000 \mu\text{mol m}^{-2} \text{s}^{-1}$  by increase of the blue-red light

311 intensity to  $950 \mu\text{mol m}^{-2} \text{s}^{-1}$ . After 60 minutes, the blue-red light  
312 intensity was switched back to  $50 \mu\text{mol m}^{-2} \text{s}^{-1}$  (total intensity  
313 including green measuring light:  $100 \mu\text{mol m}^{-2} \text{s}^{-1}$ ) for 30 minutes to  
314 trigger stomatal closure.

### 315 [Statistical Tests](#)

316 All statistical tests were performed in Origin (OriginLab). Normality  
317 tests for the distribution of fit or measured parameters were  
318 performed with Shapiro-Wilkins test. KWANOVA and paired Wilcoxon  
319 signed rank tests were used to test for significant differences between  
320 sets of parameters that were not normally distributed. Spearman's  
321 correlation was calculated between set of parameters that were not  
322 normally distributed. For normally distributed data, students t-test  
323 and Pearson correlations were used. The number of replicates and the  
324 probability scores are mentioned with the test results.

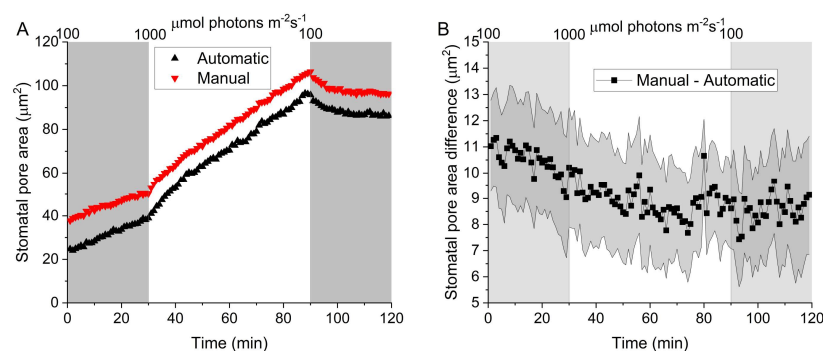
### 325 [Results](#)

326 Our portable microscope enabled the imaging of stomatal dynamics  
327 of *Chrysanthemum* in the growth environment by its selective use of  
328 green light, with limited effects on stomatal movement (Jones *et al.*,  
329 2022) for imaging at a low light intensity, while using stepwise  
330 changes in red and blue light intensity to trigger changes in stomatal  
331 aperture.

#### 332 [Comparison of stomatal pore area dynamics between 333 automatic and manual selection of best focus per pore](#)

334 The dynamics of the pore area in the focus projection images  
335 (automatic) were generally well correlated with those in the manually  
336 selected best single focus images (Fig. 4A, SM Fig. 2, table 1).  
337 However, the pore area tended to be smaller when derived  
338 automatically: pores were on average  $10 \mu\text{m}^2$  smaller in automatic  
339 than in manual images, with slightly larger differences during the first  
340 45 min of the experiment (Fig. 4B). To investigate if this difference in  
341 pore area arose from spatial dependence of image quality, we  
342 correlated the difference against positional coordinates of the  
343 stomata. We found that these differences between automatic and

344 manual-derived pore area were neither correlated to the x, y position  
 345 of the pore in the image, nor to the distance of the pore to the image  
 346 edge, nor to the position in the stack of the manual images (Table 2).  
 347 Further, pore opening in the automatic images tended to be faster  
 348 than in the manual images, as indicated by the  $k_i$  parameter of the  
 349 model (Table 1), although this difference was not significant. Pore  
 350 closing speed was similar in both image sequences. In conclusion, the  
 351 automatic method can correctly assess the kinetic parameters of the  
 352 stomata but underestimates the true pore area.



353

354 *Figure 4. A Example of kinetics of one stoma analyzed with the focus projection*  
 355 *(automatic) method or in the manually selected best single focus images (manual)*  
 356 *demonstrating their correlation. All individual stomata curve pairs are in SM Fig 2. B*  
 357 *Averaged difference in pore area between the same pore analyzed in the automatic*  
 358 *images and the Manual images in all timepoints of the shade-sun-shade transition.*  
 359 *The shaded area around the curve indicates the standard error of means. Light grey*  
 360 *blocks indicate shade periods.*

361 *Table 1. Averaged fit parameters of the kinetic model for the pore areas in the*  
 362 *automatic and manual images as well as Pearson's correlation coefficient between the*  
 363 *pore area kinetics in automatic and manual images. NS indicates no significant*  
 364 *differences between the means as tested with a paired T-test (for  $k_i$  and  $k_d$ ) or non-*  
 365 *significant correlation ( $P>0.05$ ).  $k_i$  and  $k_d$  were normally distributed (Shapiro-Wilkins*  
 366 *test). Errors indicate the SEM.*

Method (n=10)	$k_i$ (min) NS	$k_d$ (min) NS	Pearson's r	
			automatic	manual
automatic	$30 \pm 2$	$2.8 \pm 0.4$	1	$>0.97$ ( $P<0.0001$ )
manual	$36 \pm 3$	$3 \pm 1$	$>0.97$ ( $P<0.0001$ )	1

367

368 *Table 2 Pearson's or Spearman's correlation coefficient between the average*  
 369 *automatic-manual difference and the x, y and z coordinate of the pore and the*  
 370 *distance to the image edge. Distance to image edge was not normally distributed in*  
 371 *contrast to the coordinates or the z-position in the stack (Shapiro-Wilkins test)*

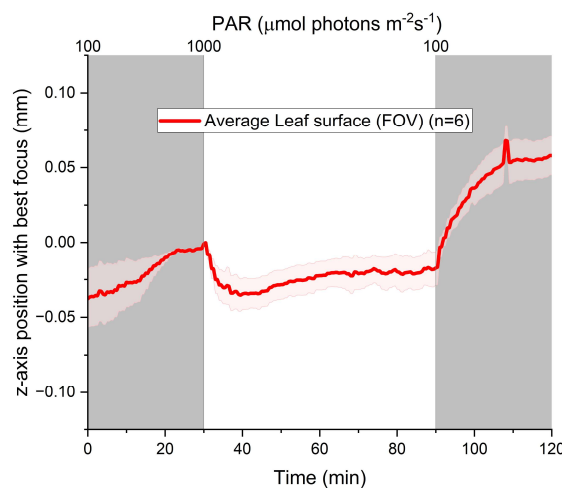
372 therefore Spearman correlation is used. NS indicates non-significant correlation  
373 ( $P > 0.05$ ).

Pearsons or Spearman's $\rho$	r	X coordinate (r)	Y coordinate (r)	Distance to image edge ( $\rho$ )	z-position in stack (r)
manual- automatic		-0.16 (NS)	0.20 (NS)	0.24 (NS)	-0.36 (NS)

374

### 375 Movement of the leaf surface upon light intensity change

376 Because we recorded the changes in the position of the leaf surface  
377 via automatic adjustments by the step motor of the microscope, we  
378 could observe the movement of the leaf surface upon changes in light  
379 intensity (Fig. 3C). Specifically, after the switch from 100 to 1000  $\mu\text{mol}$   
380  $\text{m}^{-2} \text{s}^{-1}$ , there appeared to be some shrinkage or bending of the leaf as  
381 its surface moved away from the camera, whereas upon the transition  
382 from 1000 to 100  $\mu\text{mol} \text{m}^{-2} \text{s}^{-1}$ , the opposite happened (Fig. 5). These  
383 leaf surface movements had no impact on imaging, because the FOV  
384 of the leaf was always captured in focus within the stack. Leaf surface  
385 movements resulted in a minor displacement of the position of the  
386 leaf surface in the stack (SM Fig. 3). In conclusion, the active  
387 repositioning of z-positions of each new image stack based on the z-  
388 position of the leaf surface in the previous image stack, is necessary  
389 because the leaf surface moves hundreds of micrometers during light  
390 intensity changes.



391



392 *Figure 5. Position on the z-axis of the imaged area of the leaf surface where the largest*  
393 *area of leaf surface was in focus during the shade-sun-shade experiment. The red*  
394 *curve indicates the average of six biological replicates, and the shaded area indicates*  
395 *the standard error of the mean. The z-axis position at t=30 min was set to zero for*  
396 *each measurement before averaging. The Focus position (y-axis) indicates the middle*  
397 *position of the 100 images per one minute timepoint that were spaced 1  $\mu\text{m}$  apart.*

398 Stomatal dynamics of *Chrysanthemum* leaves during a  
399 shade-sun-shade transition.  
400

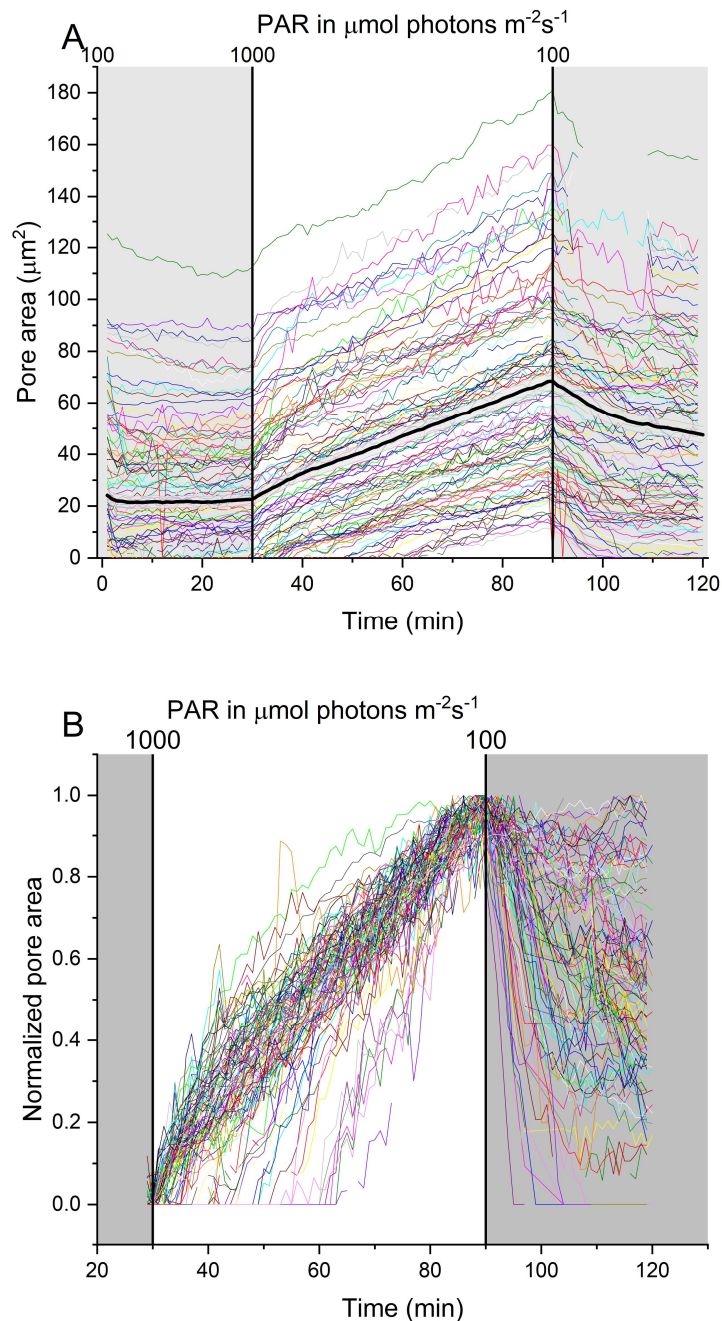
401 To further demonstrate the power of our method, we continued to  
402 test the resolving power of the method in leaves of six plants split over  
403 two cultivation periods. After image processing, ~78% (124/158) of  
404 the stomata in the FOV were kinetically resolved (Fig. 6).

405 While overall stomatal aperture was in a steady-state in the  
406 initial shade condition (Figure 6, 0-30 min), individual stomata  
407 showed minor movements by either opening or closing only slightly.  
408 After the light intensity was increased, 62% of stomata (77)  
409 immediately opened their pores, while the rest showed a lag in their  
410 response of up to 40 min (Fig. 6). A duration of 60 min at 1000  $\mu\text{mol}$   
411  $\text{photons m}^{-2} \text{s}^{-1}$  was not sufficient to reach a steady-state pore opening  
412 for most stomata, which led to more uncertainty in the fitted  $k_i$   
413 parameter than if a steady-state had been reached. Figure 7 and 8  
414 show the distribution of the fitted parameters in violin plots.

415 The light intensity decrease led to an immediate closing  
416 response of thirteen stomata (13%), with the rest lagging in their  
417 response. Nineteen stomata did not close at all within the 30 min sun-  
418 shade transition (20%). These 19 stomata had a significantly higher  
419 steady-state aperture under high light intensity than the 78 stomata  
420 that closed during the shade phase ( $A_{\text{max-l}} 78 \pm 6 > 60 \pm 3$ , mean+SE,  
421  $P < 0.02$ , KWANOVA). In addition, they were not evenly distributed  
422 among the biological replicates: of the 19 non-closing stomata, 15  
423 were located on one biological replicate and the remaining four  
424 stomata from two others (1 and 3 per replicate), while the remaining  
425 replicates had none. A steady-state was reached for all but five of the  
426 stomata that closed within the 30 min sun-shade transition, however  
427 this new steady-state pore aperture at lower light intensity was larger

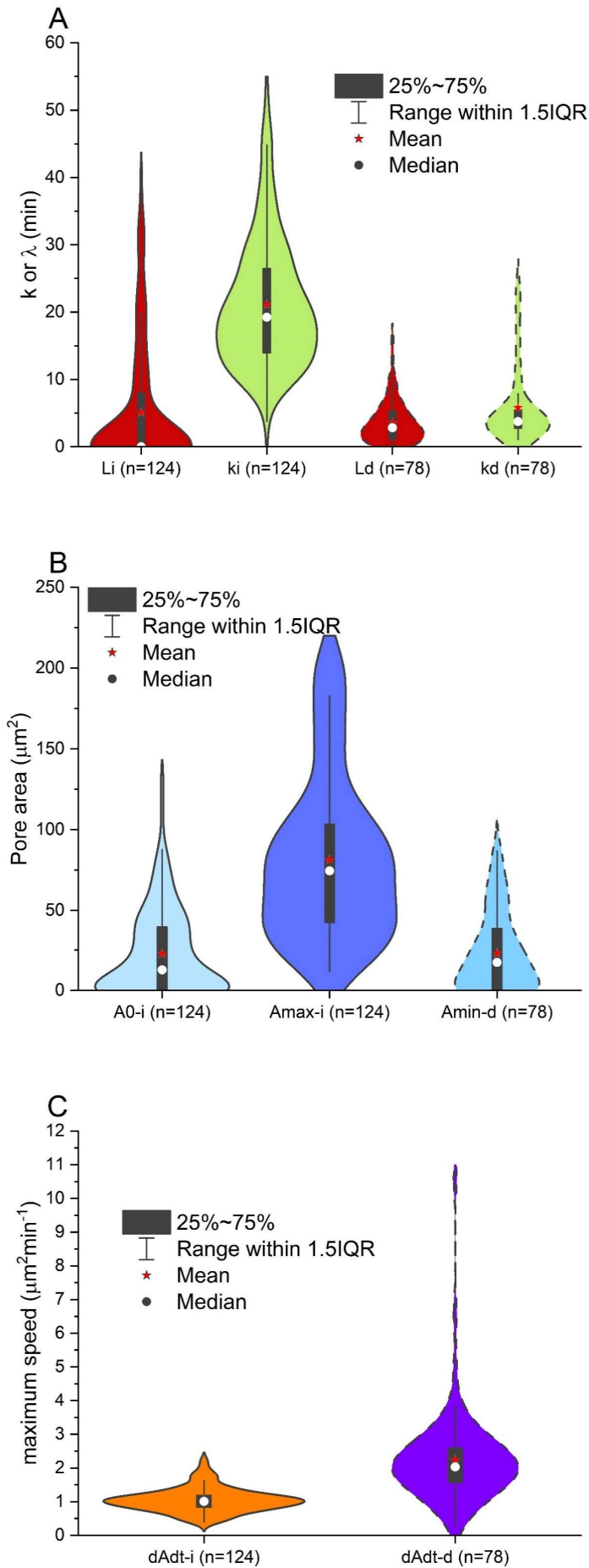


428 than the initial steady-state level of these stomata before the opening  
429 light stimulus was applied ( $A_{\text{min-d}} 23 \pm 3 \mu\text{m}^2 > A_{0-i} 13 \pm 2 \mu\text{m}^2$ , mean+SE,  
430 paired sample Wilcoxon Signed Ranks Test,  $P < 0.005$ ).

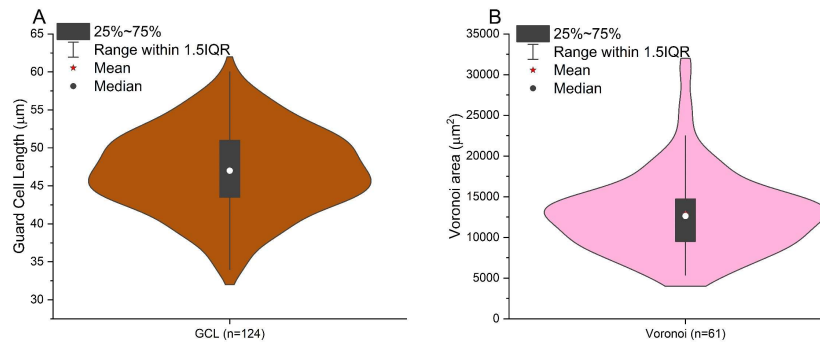


431

432 *Figure 6. A Example of stomatal pore area changes upon a shade-sun-shade*  
433 *transition. Single lines show values of 124 individual stomatal pores from six*  
434 *chrysanthemum plants. The thick black curve represents the average, and the grey*  
435 *area the standard error of the mean. During the second shade period, a temporary*  
436 *loss of focus during one measurement caused fewer data points to be recorded. B*  
437 *Kinetics of all stomata shown in A, normalized to 0 at  $t=29$  min just before the light*  
438 *intensity was increased and to 1 for the maximum aperture reached at  $1000 \mu\text{mol}$*   
439 *photons  $\text{m}^{-2}\text{s}^{-1}$ , highlighting the differences in opening lag ( $\lambda_i$ ) and stomatal closure.*



441 *Figure 7. Distributions of parameters obtained from model fits to stomatal kinetics*  
442 *(raw data shown in Fig. 6). A. Opening and closing lag ( $\lambda_i$ ,  $\lambda_d$ ) as well as time constants*  
443 *for opening and closing ( $k_i$ ,  $k_d$ ) B. Steady-state stomatal pore area in the initial shade*  
444 *period  $A_0$ , in sun period  $A_{max-i}$  and the final shade period  $A_{min-d}$ . C Maximum opening*  
445  *$dAdt$ , and maximum closing speed  $dAdt$ . Red stars represent the mean and circles the*  
446 *median of the data, 25-75% indicates the position where the middle 50% of the values*  
447 *in the distribution can be found and the range within 1.5 IQR indicates which part of*  
448 *the distribution are not considered to be outliers.*

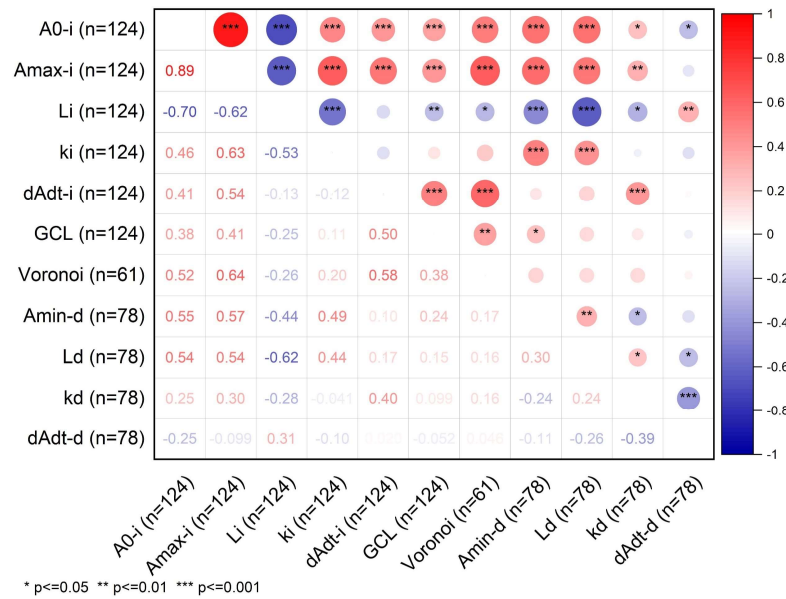


449

450 *Figure 8 Distribution of A. guard cell length (GCL) of the stomata in Fig. 6 measured*  
451 *in the last image during the high light period ( $t=89$  min) and B. Voronoi area*  
452 *representing the area on the leaf that is within the shortest distance to a given stoma.*  
453 *Red stars represent the mean and circles the median of the data, 25-75% indicates*  
454 *the position where the middle 50% of the values in the distribution can be found and*  
455 *the range within 1.5 IQR indicates which part of the distribution are not considered to*  
456 *be outliers.*

457 Of the 78 stomata that opened and closed, only six  
458 had a larger  $k_i$  than  $k_d$  and those  $k_d$  were the very highest among the  
459 78 values (Fig. 7A), meaning that for the majority of stomata, the  
460 steady-state when closing was reached faster than when opening.

461 Of all parameters, only guard cell length was normally distributed  
462 (Shapiro Wilkins test,  $p < 0.001$ ), therefore Spearman's nonparametric  
463  $\rho$  was used to investigate correlations. A higher aperture during the  
464 first shade phase ( $A_{0-i}$ ) was strongly correlated with a higher maximum  
465 aperture in the sun phase (Figure ,  $A_{max-i}$ ,  $\rho = 0.89$ ,  $p < 0.001$ ) and  
466 moderately correlated to the steady-state aperture during the second  
467 shade phase ( $A_{min-d}$ ,  $\rho = 0.55$ ,  $p < 0.001$ ).  $A_{max-i}$  was also moderately  
468 correlated to  $k_i$  ( $\rho = 0.63$ ,  $p < 0.001$ ) and  $dAdt_{max-i}$  open ( $\rho = 0.54$ ,  
469  $p < 0.001$ ). The time-lag of the opening response ( $\lambda_i$ ) was negatively  
470 correlated to  $A_{0-i}$ ,  $A_{max-i}$ ,  $k_i$ , GCL, Voronoi  $A_{min-d}$ ,  $k_d$  and  $\lambda_d$ , and positively  
471 correlated to the closing speed  $dAdt_{max-d}$ . Interestingly, the time-lag of  
472 the closing response ( $\lambda_d$ ) was correlated to  $A_0$  ( $\rho = 0.54$ ,  $p < 0.001$ ) and  
473  $A_{max-i}$  ( $\rho = 0.54$ ,  $p < 0.001$ ).



474

475 *Figure 9. Spearman's correlation matrix between parameters of individual*  
 476 *Chrysanthemum stomata (based on Fig. 7 and 8).  $A_0$ , fitted steady-state aperture*  
 477 *under 100  $\mu\text{mol photons m}^{-2}\text{s}^{-1}$  PPFD.  $A_{max}$ , predicted steady-state aperture*  
 478 *under 1000  $\mu\text{mol photons m}^{-2}\text{s}^{-1}$  PPFD;  $k_i$ , time constant for apertures to increase to  $A_{max}$*   
 479 *under 1000  $\mu\text{mol photons m}^{-2}\text{s}^{-1}$  PPFD;  $k_d$ , decrease from  $A_{max}$  to  $A_f$  under 100  $\mu\text{mol}$*   
 480 *photons  $\text{m}^{-2}\text{s}^{-1}$ ;  $l$ , initial lag in the response time of stomatal aperture to a step*  
 481 *increase in PPFD; max dAdt, maximum rate of stomatal opening or closing to an*  
 482 *increase or decrease in PPFD from 100 to 1000  $\mu\text{mol photons m}^{-2}\text{s}^{-1}$  or vice versa.*  
 483 *Anatomical parameters of guard cell length (GCL) and width (GCW) were also*  
 484 *compared. The Voronoi area represents the fraction of the leaf that is within the*  
 485 *shortest distance to that stoma.*

486 Among the anatomical parameters, Guard cell length (GCL)  
 487 was moderately correlated to the maximum opening speed ( $dAdt_{max-i}$ ,  
 488  $\rho=0.5$ ,  $P<0.001$ ), initial ( $A_{0-i}$ ,  $\rho=0.38$ ,  $P<0.001$ ), maximum ( $A_{max-i}$ ,  
 489  $\rho=0.41$ ,  $P<0.001$ ) and final aperture ( $A_{min-d}$ ,  $\rho=0.24$ ,  $P<0.05$ ) as well as  
 490 the Voronoi area ( $\rho=0.38$ ,  $P<0.01$ ). The Voronoi area was in addition  
 491 moderately correlated with  $A_{0-i}$  ( $\rho=0.52$ ,  $p<0.001$ ),  $A_{max-i}$  ( $\rho=0.64$ ,  
 492  $p<0.001$ ) and the maximum opening speed ( $dAdt_{max-i}$ ,  $\rho=0.58$ ,  
 493  $P<0.001$ ).

494 In conclusion, our method resolved 11 different kinetic and  
 495 non-kinetic parameters for up to 78% of the stomata within the FOV  
 496 to generate new insight into the WUE of individual stomata.

## 497 Discussion

498 The portable microscope presented here provides a new way  
 499 to study the opening and closure of individual stomata in the growth

500 environment. This method facilitates analyses of individual stomatal  
501 dynamics in relation to their local morphology and anatomy.

502 Our hardware and method are distinct from several previous  
503 studies that imaged stomata *in situ* (Kaiser & Kappen, 1997, 2001;  
504 Kaiser, 2009; Grantz *et al.*, 2018). The main differences and their  
505 advantages and disadvantages are listed in Table 3.

506 *Table 3 Differences between this work and prior studies (Kaiser & Kappen, 1997,*  
507 *2001; Kaiser, 2009; Grantz et al., 2018) with their advantages and disadvantages*

Difference	Advantage	Disadvantage	Study
525 nm versus 880 nm imaging light	Higher optical resolution	Imaging light is photosynthetically active	(Kaiser & Kappen, 1997, 2001)
Autofocus per stack instead of per pore	Imaging speed is not limited by number of stomata but only by stack size	Can cover stomata in only a small area of the leaf (~1 mm <sup>2</sup> )	(Kaiser & Kappen, 1997, 2001; Kaiser, 2009; Grantz et al., 2018)
Kinetic model fitting	Quantitative comparison of fit parameters	-	
No Gas exchange chamber	More natural boundary layer conditions	No information on leaf bulk gas exchange	(Kaiser & Kappen, 1997, 2001; Kaiser, 2009; Grantz et al., 2018)

508

509 A major contrast with prior studies is that the way the leaf is clamped  
510 was minimally invasive to the leaf's microenvironment, with free  
511 transpiration on both sides of the leaf, because no gas exchange  
512 measurements were performed. The boundary layer was not  
513 regulated, reflecting a more natural condition. Another major  
514 difference was the sampling of stomata: prior studies have randomly

515 sampled stomata distributed over a large area ( $>1 \text{ cm}^2$ ), whereas we  
516 resolved as many stomatal kinetics as possible within  $\sim 1 \text{ mm}^2$ , to  
517 relate their kinetic behavior to the local morphology on the surface of  
518 the leaf. In doing so, we reached a higher time resolution of, on  
519 average, 21 stomata  $\text{min}^{-1}$  compared to 75 stomata once every 6-7  
520 min (Grantz *et al.*, 2018). Another major difference is that we fit a  
521 kinetic model to individual stomatal aperture kinetics to investigate  
522 the relation of model parameters to parameters related to the local  
523 morphology on the leaf surface.

524 A clear limitation of our method is the underestimation of pore area  
525 compared to manual selection of the best focus per pore (Fig. 3). This  
526 is especially concerning for pores that only opened minimally, as  
527 these were likely found to be closed even when they were not actually  
528 closed. Such pores are scored with large opening lags that are  
529 erroneous. The minimal resolvable pore area is effectively limited  
530 beyond the optical resolution of the microscope to  $\sim 10 \mu\text{m}^2$  for the  
531 20x objective used here. This limitation of the automated focus  
532 projection method is not absolute but scales with the numerical  
533 aperture of the objective. Minimally resolvable pore area can thus still  
534 be improved at the cost of FOV size and working distance. The  
535 smaller pore area in the automated focus projection method is likely  
536 due to differences in absolute contrast between the automatic and  
537 the manual images that lead to different thresholding by the  
538 percentile method. Implementation of the state-of-the art in  
539 computer vision solutions, such as StomaAI (Sai *et al.*, 2023), for  
540 stomatal measurements and segmentation may increase the  
541 resolving power in the future.

542           The analysis of dozens of *Chrysanthemum* stomata led us to  
543 the novel insight that a larger aperture at steady-state in high light  
544 intensity ( $A_{\text{max-i}}$ ) was positively correlated with the time lag for the  
545 closing response ( $\lambda_d$ ), with a large negative consequence for WUE  
546 (Lawson & Blatt, 2014). Stomata with larger pores may suffer from a  
547 larger lag in their response because of a stronger hysteresis.  
548 Hysteresis is related to the opening speed at the time of the change



549 in light intensity, with opening speed ( $dA/dt_i$ ) positively correlated with  
550 the maximum aperture ( $A_{\max-i}$ ).

551           Across a number of species, Deans *et al* (Deans *et al.*, 2019)  
552 found that a large stomatal conductance at high light intensity was  
553 correlated with fast closing response, to compensate in water-use  
554 efficiency. We could not corroborate this for  $A_{\max-i}$  and their rate of  
555 closing  $dA/dt_{\max-d}$  in individual stomata. *Chrysanthemum* stomata  
556 with a large aperture thus have a disproportionately large negative  
557 contribution on WUE normalized for their apertures, due to the  
558 observed correlation between  $A_{\max-i}$  and the time lag for the closing  
559 response  $\lambda_i$ , combined with a similar closing speed as stomata with a  
560 smaller aperture.

561           Stomatal size was found to be negatively correlated with WUE  
562 across different *Arabidopsis* ecotypes (Dittberner *et al.*, 2018). A high  
563 WUE is typically associated with a fast rate of stomatal closure  
564 (Lawson & Vialet-Chabrand, 2019), and smaller stomata have indeed  
565 often been observed to open or close faster than larger stomata  
566 (Drake *et al.*, 2013; Kardiman & Ræbild, 2018; Durand *et al.*, 2019). In  
567 contrast, we found a strong positive correlation between GCL and  
568 maximum opening speed, as observed before for elliptical/kidney-  
569 shaped stomata (McAusland *et al.*, 2016). The relation between  
570 stomatal size and speed may thus be more complex than previously  
571 assumed. This relation could be species dependent, as no correlation  
572 between stomatal size and speed was found in a diverse range of  
573 plants with differing stomatal morphologies and physiological  
574 behaviors (Haworth *et al.*, 2018). Additionally, it may depend on the  
575 type of stimulus, since e.g. step changes in light intensity and in VPD  
576 often trigger different dynamics (Durand *et al.*, 2019).

577           The opening and closing speeds of stomata, that were not  
578 correlated here, were found to be positively correlated in some  
579 studies (Haworth *et al.*, 2021), but showed no correlation in others  
580 (Xiong *et al.*, 2018). The correlation between opening and closing  
581 speed may thus be species and/or condition dependent.



582           Mesophyll airspace formation is linked to functional pores  
583 (Lundgren *et al.*, 2019), therefore we assumed that there was a  
584 mesophyll airspace under each functional pore. We further assumed  
585 that the diffusion pathway to the surrounding mesophyll was  
586 approximately equal in all directions and that the majority of CO<sub>2</sub>  
587 feeding the mesophyll directly underneath the leaf surface was  
588 supplied by stomata on that surface and not by stomata on the  
589 opposite side of the leaf. Finally, we assumed that airspace and  
590 mesophyll were homogeneously distributed below the pores. The  
591 positive correlation of A<sub>0</sub> and A<sub>max</sub> with the Voronoi area of the stoma  
592 can be interpreted as the stoma matching its aperture to the local  
593 demand for CO<sub>2</sub> from the direct mesophyll below. If a stoma has to  
594 supply a larger area of mesophyll, its steady-state aperture will be  
595 larger under all light conditions than that of a stoma that supplies a  
596 smaller area of mesophyll. The same reasoning can be applied to the  
597 opening speed: if one stoma ‘feeds’ a large mesophyll area with CO<sub>2</sub>,  
598 drawdown of C<sub>i</sub> during photosynthetic induction will likely be stronger  
599 compared to a stoma that supplies a smaller area, and the signal for  
600 fast stomatal opening may thus be stronger. An alternative  
601 explanation for the positive correlation of A<sub>0</sub> and A<sub>max</sub> with the  
602 Voronoi area may be that a larger Voronoi area supplies the guard  
603 cells with more solutes, water and space to open than a smaller  
604 Voronoi area.

## 605 [Conclusions](#)

606 We showed here that *in vivo* microscopy of stomata in the leaf’s  
607 growth environment could be used to generate stomatal opening and  
608 closure dynamics in dozens of neighboring individual stomata. Our  
609 method, which uses green light for imaging and generates focus  
610 projections of the entire leaf surface within the FOV, resolved the  
611 kinetics of 78% of *Chrysanthemum* stomata, on average 21 stomata  
612 min<sup>-1</sup> during a shade-sun-shade transition. Pore area kinetics were  
613 fitted with a model developed for stomatal conductance time courses  
614 and show the substantial variation between individual stomata.  
615 Correlation between the fitted parameters led us to discover that

616 pores with larger apertures in high light have a larger lag time in their  
617 closing response, and thereby contribute disproportionately to a  
618 lower WUE.

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## 624 Competing interests

625 None declared

## 626 Author contributions

627 Conceptualization T.B. and J. S., Hardware T.B. R.S. Software T.B.,  
628 R.S., Growing facilities E.K., Experimentation T.B., Data Analyses T.B.,  
629 Writing the first draft T.B., Critical review of the manuscript T.B., E.K.,  
630 J.S., Funding J.S.

## 631 Data availability

632 All data is available upon request. For more information and  
633 software updates: [GitHub - Plantenna/Stomata-microscope: Non-invasive microscopic imaging of individual stomatal kinetics in the growth environment with high resolution](#)  
634  
635

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## 808     Supporting Information

809     SM Fig 1. Example of stomatal aperture dynamics of a single stoma  
810     and the model fit for opening (t=30-90) and closing (t=90-120). Raw  
811     aperture data is in black and the model fit curve in red.

812     SM Fig. 2 Stomatal aperture dynamics of focus projection images  
813     (automatic) and manually determined best single focus images  
814     (manual) demonstrating their correlation.

815     SM Fig. 3 Example of how the position of the best focus for a stoma  
816     can change in a stack during the shade-sun-shade experiment due to  
817     the movement of the leaf surface (Fig. 5).

818     SM Movie 1 Example of a processed z-stack (100images) of a single  
819     timepoint in the experiment. Stomata positions are indicated by red  
820     ovals.

821     SM movie 2 Example of the processed time stack (video) of a full  
822     experiment (119 focus projections of the z-stacks of each timepoint).  
823     Stomata positions are indicated by red ovals.

824 SM video 3 Example of the processed time stack (video) of a full  
825 experiment, zoomed in on a single stoma.

826 SM video 4 The same video of 3 after thresholding.

827 SM video 5 The same video as 3 and 4 after segmentation.

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