

Measuring the diurnal pattern of leaf hyponasty and growth in Arabidopsis – a novel phenotyping approach using laser scanning

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1	Measuring the diurnal pattern of leaf hyponasty and growth in		
2	Arabidopsis – a novel phenotyping approach using laser		
3	scanning		
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10	Summary (50 to 80 words, three sentences for non Scientists)		
11	Increased leaf elevation angle (hyponasty) and leaf elongation in Arabidopsis are		
12	caused by different environmental stimuli such as shading by surrounding vegetation.		
13	Here we report on a phenotyping approach based on laser scanning to measure the		
14	diurnal pattern of these two growth responses. Individual plants can be monitored		
15	during several days under different light conditions, with high temporal resolution, in		
16	high-throughput and non-invasively.		
17			
18	Abstract (200)		
19	Plants forming a rosette during their juvenile growth phase, such as Arabidopsis		
20	thaliana, are able to adjust size, position and orientation of their leaves. These growth		
21	responses are under the control of the plants circadian clock and follow characteristic		
22	diurnal rhythms. For instance, increased leaf elongation and hyponasty -defined here as		
23	the increase in leaf elevation angle- can be observed when plants are shaded. Shading is		
24	either caused by a decrease in the fluence rate of photosynthetically active radiation		
25	(direct shade) or a decrease in the fluence rate of red compared to far-red radiation		
26	(neighbor detection).		
27	In this paper we report on a phenotyping approach based on laser scanning to measure		
28	the diurnal pattern of leaf hyponasty and increase in rosette radius. In short days, leaves		
29	showed a constitutively increased leaf elevation angles compared to long days, but the		
30	overall diurnal pattern and the magnitude of up and downward leaf movement was		

31 independent of day length. Shade treatment led to elevated leaf angles during the first

1 day of application, but did not affect the magnitude of up and downward leaf movement 2 in the following day. 3 Using our phenotyping device, individual plants can be monitored during several days 4 under different light conditions, with low temporal resolution, in high-throughput and 5 non-invasively. It hence represents a proper tool to phenotype light- and circadian 6 clock-mediated growth responses in order to understand the underlying regulatory 7 genetic network. 8 9 10 Key words 11 Arabidopsis, phenotyping, laser scanning, image processing, hyponasty, petiole angle, 12 leaf elongation, diurnal, circadian clock 13 14 15 Introduction 16 Being sessile, higher plants need to adapt to their ever-changing environment to 17 optimize primary production and warrant good conditions for their offspring. A way to 18 achieve this is through phenotypic plasticity of the aerial plant body in response to 19 stimuli such as day length, light intensity, solar angle, water/nutrient availability or 20 temperature (de Kroon et al. 2009; Novoplansky 2009; Sultan 2010). Adjusting leaf 21 angle is an important aspect of plant plasticity that optimizes the interception of 22 incoming light (Ehleringer 1988) or avoids temperature-related stress (Medina et al. 23 1978). 24 Plants forming a rosette during their juvenile growth phase, such as Arabidopsis 25 *thaliana*, are able to alter the position, orientation, length, width and shape of their 26 leaves as well. Numerous stimuli can cause these growth responses such as radiation, 27 touch, chemicals, gravity, water or temperature. For instance an upward movement of

- 28 leaves (referred to as leaf hyponasty), which is the increase in the leaf elevation angle,
- can be observed in response to submergence (Cox et al. 2003; Voesenek et al. 2006)
- 30 or increased temperatures (Koini *et al.* 2009). Hyponasty enables plants to position
- 31 their leaves above the water surface in case of submergence or to increase transpiration
- 32 for cooling in case of high temperature. Changes in light quantity, quality or direction can

also trigger leaf movements and increased leaf elongation. Arabidopsis leaf blades are 1 2 usually oriented perpendicular to the incoming light (phototropism). A decrease in the fluence rate of photosynthetic active radiation (E_{PAR}) by direct shading leads to leaf 3 hyponasty and increased petiole elongation (Keller et al. 2011) or hypocotyl elongation 4 5 (Keuskamp *et al.* 2011). This shade avoidance response is not only triggered by a low E_{PAR} , but also by a decrease in the fluence rate of red (R) compared to far-red (FR) light 6 (Mullen et al. 2006; Sasidharan et al. 2010). The reduced R-FR ratio arises from an 7 8 increased scattering of FR radiation by surrounding vegetation and thus indicates the 9 presence of neighbor plants, which constitute potential future competitors. Leaf hyponasty and increased petiole elongation in response to low E_{PAR} and low R-FR ratio 10 11 aim at positioning leaves in the upper part of the canopy, where light conditions are 12 more favorable. In addition to external stimuli, leaf hyponasty and elongation also depend on the internal circadian clock (Millar et al. 1995; Salter et al. 2003). In 13 14 Arabidopsis grown in regular day-night cycles, leaves move and grow according to a 15 rhythmic diurnal pattern. Leaf growth rate is highest in the early morning and rapidly 16 declines thereafter (Wiese et al. 2007). The leaf elevation angle increases during the day followed by a decrease at the end of the night before dawn (Millenaar et al. 2005; 17 18 Mullen et al. 2006; Sasidharan et al. 2010). 19 In spite of these dynamic aspects of plant growth, measurements of leaf or petiole 20 elevation angle and length in Arabidopsis are quite challenging. Most studies have relied on ruler (Mullen et al. 2006), protractor (Keller et al. 2011), or photogrammetry 21 22 (Millenaar et al. 2005) as measurement technique. However, plants need to be 23 sampled or illuminated with visible light applying these methods, which perturbs the plants internal circadian clock, and does not allow monitoring of diurnal leaf 24 growth over several days. Our approach based on laser-scanning and infra-red 25 photography allows to over come this obstacle. In addition, laser-scanning as three-26 27 dimensional (3D) imaging technique allows to non-destructively measure the whole 28 3D surface of leaves and opens up exciting possibilities to study architectural traits of different leaves during several days. 3D imaging techniques have already been 29

30 applied in plant phenotpying, including laser scanning (Kaminuma et al. 2004),

- 31 phase-shifting projected fringe profilometry (Dornbusch et al. 2007), X-ray
- 32 computed tomography (Kaminuma *et al.* 2008; Dhondt *et al.* 2010) or optical projection tomography (Lee *et al.* 2006). The most

1 advanced approach for Arabidopsis phenotyping using these techniques has been 2 developed by Kaminuma et al. (2004) who proposed to use laser scanning to extract 3 morphological traits of leaves from measured 3D data. They succeeded in measuring 4 traits such as the orientation of the blade surface and epinasty in a detailed and 5 automated way. Measurements were done on Arabidopsis plantlets with two true leaf 6 blades (plus two cotyledons), where leaves were almost horizontally oriented and did 7 not overlap. However, the architecture of Arabidopsis plants is usually more complex. 8 Leaves can be oriented more vertically and overlap in later growth stages and are 9 thus not fully captured by the laser scanner. Hence an automated 3D image analysis 10 does not seem applicable. Indeed, to our knowledge, there has been no follow-up 11 work reported using the approach by Kaminuma and co-workers (2004) so far. 12 Here, we take up the laser scanning approach and propose a phenotyping pipeline using 13 laser scanning and subsequent image analysis. Our goal is to extract traits from these 14 images that can be reliably linked to leaf elevation angle and plant size to measure these 15 growth responses. In this paper, we present our phenotyping pipeline, which includes: i) 16 technical specifications of the ScanAlyzer HTS, ii) plant growth conditions (mainly 17 light and daylength) and iii) and image processing algorithm implemented in Matlab (The MathWorks Inc., Natick, USA). We furthermore give examples of possible 18 19 applications of our phenotyping method by presenting data on diurnal leaf hyponastic responses caused by different day-length, light intensities (EPAR), light quality (R-FR 20 21 ratio) and genetic background.

22

23 Materials and Methods

24 Plant growth

25 The Arabidopsis thaliana accession Columbia (Col-0) and the sav3-2 mutant (Tao et al. 26 2008) were used. Seeds were stratified three days on moist filter paper in the dark at 27 4°C and sown on a mixture (50:50) of peat-rich soil and vermiculite. Deionized water 28 was added to saturate the growth medium. The soil surface around plants was covered 29 with a thin layer of charcoal to minimize laser reflection. Prior to scanning, plants were 30 kept in a Percival CU-36L4 incubator (Percival Scientific, Inc., Perry, USA) at 21°C for 31 15 days at long day (16h photoperiod) or 20 days at short day (8h photoperiod). Plants 32 were transferred to the ScanAlyzer HTS 24h before scanning at growth stage 1.05

1 (Boyes *et al.* 2001) maintaining the entrained photoperiod. The device is 2 Fluorescence tubes were 100 cm above plant level and emitted cool white light (light 3 color 865) at a fluence rate $E_{PAR} = 180 \,\mu\text{mol m}^{-2} \,\text{s}^{-1}$ in the incubator.

4

5 Light conditions

6 The spectral composition of the light in the ScanAlyzer HTS (Lemnatec GmbH, 7 Würselen, Germany) and the Percival incubator was measured using a USB2000+ UV-8 VIS spectrometer (Ocean Optics, Inc., Dunedin, USA). For each measurement, 50 9 spectra were recorded at plant level and different horizontal positions, and averaged to 10 a mean spectrum. The fluence rate of photosynthetically active radiation (E_{PAR}) was 11 computed as the integral from 400nm to 700nm. Likewise, the fluence rate of far-red 12 radiation ($E_{\rm FR}$) was calculated by integrating the spectrum from 680nm to 760nm (see 13 Supplemental Figure 1). Four different light setups were used in our experiments (Tab. 14 1). In the ScanAlyzer HTS, E_{PAR} was decreased installing a gray plexiglas filter panel 6 15 cm underneath the fluorescence tubes. Likewise, $E_{\rm FR}$ was altered switching on the 16 dimmable built-in FR diodes. Different light treatments did neither influence 17 temperature nor humidity inside the measurement chamber due to its large volume (~ 2 18 m^{3}) and high air-exchange rate with the surrounding air-conditioned room (21°C, 70%)

19 rel. humidity).

20

21

22 Technical specifications of the ScanAlyzer HTS

23 Our custom-built version of the Scanalyzer HTS is equipped with fluorescence tubes

- 24 providing cool white light (light color 865) and dimmable FR diodes ($\lambda = 740$ nm). The
- 25 imaging unit entails a NIR Laser ($\lambda = 905$ nm, $P \approx 10$ mW) pointing nadir, a camera to
- 26 record the laser beam Vosskühler CMC-1300 (Allied Vision Technologies GmbH,
- 27 Stadtroda, Germany) mounted 45° in relation to the laser and a further charged coupled
- 28 device camera (CCD) camera (Basler scA1400-17gc, Basler AG, Ahrensburg,
- 29 Germany) mounted nadir to record RGB images (Fig. 1a). The CCD camera is equipped
- 30 with a ring of near infrared (NIR) diodes ($\lambda = 940$ nm) around the lens to record images
- 31 in darkness.

1 The imaging unit can be programmed to move to specific locations within the 2 measurement chamber approximated 15 cm above plant level. Images can be recorded 3 at a specified times or time-intervals. The measurement space in the ScanAlyzer HTS is 4 divided into 6 slots. Each slot holds a tray containing a maximum of 12 pots (12.3cm-5 by-7.9cm in size) allowing a maximum imaging capacity is 72 pots. A full scan of 72 6 pots (laser scan + RGB image) takes approximately 80 minutes. For the data presented 7 here we have usually scanned less 12-36 pots per experiments at a time interval of 40 or 8 60 minutes.

9

10 Conversion of laser scanner images to 3D point clouds

11 The output of the laser scanner is a 2.5D height-scaled image, 1024-by-1728 pixels in

size (Fig. 1b). Each pixel represents a measured point. In the image, the column index

13 (*i*) is linked to the *x*-coordinate and the row index (*j*) to the *y*-coordinate of a measured

14 point. The RGB color triplet (c) of each pixel encodes its distance from the reference

15 plane (z) and thus represents the z-coordinate of a measured point (Fig. 1b).

16 In image processing, *i* and *j* are normally multiplied with a scale factor to obtain the *x*

17 and *y*-coordinate. In our case, images have a perspective distortion caused by the shape

18 of the lens and the focal length. Therefore, each [i, j] doublet has a specific x- and y-

19 coordinate, which are assigned using two distinct look-up tables (matrices). In 2.5D

20 images, the distortion of the x-y plane in the image even depends on z. Therefore we

21 need to assign an x- and an y-coordinates to each [i, j, c] triplet. The corresponding

22 look-up tables *Kx* and *Ky* are hence 3D matrices (see Supplemental Material).

To obtain *Kx* and *Ky*, a black panel (12.3cm-by-7.9cm) with a rectangular grid (1cm

spaced and 1mm wide grid lines) was scanned at seven positions ($0 \text{cm} \le z \le 6.58 \text{cm}$).

25 Intersections of grid lines in all seven scanned images were manually selected. Thus for

26 672 [i, j, c] triplets we obtained a x-coordinate and a y-coordinate, respectively, which

27 yielded a sparse matrix for *Kx* and *Ky*. The full matrices were computed by linear

28 interpolation. Scanning the grid at each *z*, we obtained seven mean values for the color

index c and the corresponding z-coordinate. These values were used to fit a second

30 order polynomial to obtain a look-up table, in which for each c is related to z.

31 The maximum dimensions of the calibrated measurement space are 13.56cm x 8.92cm x

32 6.58cm. The described algorithm was implemented in Matlab R2009b (MathWorks

- Inc., Natick, USA). Source code to convert images to point clouds, the matrices *Kx* and
 Ky and a sample image (Fig. 1b) are given in the Supplemental material.
- 3

4 Test of image conversion

- 5 First, we tested whether the proposed image conversion of pixel coordinates in 2.5D
- 6 height-scaled images (i, j, c) to Carthesian coordinates (x, y, z) was precise enough. To
- 7 do this, we scanned a panel of 23 circles at five different heights (z) and three different
- 8 inclination angles (Fig. 2a,b). Circles had a predefined area $(0.0020 \text{ cm}^2 \le A_{\text{org}} \le$
- 9 3.46cm²). Background discrimination in binary images was done by choosing a gray
- 10 threshold of 2 ($0 \le \text{gray} \le 255$). Hence all pixels with gray>0 are belong to measured
- 11 circles. In the following cluster analysis (Matlab function bwlabel), 23 clusters of
- 12 interconnected pixels were detected, each corresponding to a scanned circle. Boundary
- 13 pixels for each cluster were identified and pixel coordinates transformed to Cartesian
- 14 coordinates using the matrices *Kx* and *Ky*. Points computed this way were connected to
- 15 triangles using Delaunay triangulation.
- 16 The sum of the triangle area corresponded to the measured area of the circle A_{scan} .
- 17 Comparing measured circle area A_{scan} to predefined values A_{org} allows to demonstrate
- 18 the precision of our image conversion. For circles scanned at different height or
- 19 inclination (Fig. 4c), values for A_{scan} deviated little from A_{org} with a minor
- 20 overestimation of A_{scan} . All points fell close to the optimal 1:1 line. The mean relative
- 21 error (MRE) was 0.094 was relatively, which could be attributed to the improper area
- 22 estimation of very small circles. For circles with a size similar to Arabidopsis leaves (A
- $23 > 0.5 \text{ cm}^2$), the MRE was 0.038.
- 24

25 Processing of 3D point clouds

First, we segmented the whole measured point cloud (PC) into *m* subsets PC_{plant} that include all points measured for each plant. First we selected P_0 for each plant, which is geometrically the position of the shoot apical meristem (Fig. 3a). For time-lapse images, the selection of P_0 was done once per day for each plant at zeitgeber time t = [3, 27, 51]. For PCs measured in between two manual selections (*e.g.* 3 > t > 27), P_0 was computed using linear interpolation. Next, P_0 was subtracted from PC, shifting the origin of the

32 coordinate system of PC to P_0 . Points were assigned to PC_{plant}, which had a distance $r \leq$

1.8cm to P_0 and z > -0.1cm (see Fig. 3a). In our experiment leaves were always shorter 1 2 than the chosen value for r (Fig. 3d). The distribution of the spherical coordinates azimuth angle (θ), elevation angle (ϕ) 3 4 and radius (r) of PC_{plant} is given in Fig 3c-d. They are related to plant traits as follows: 5 θ = phylotaxis, φ =leaf elevation angle and *r* = rosette radius. 6 7 Data processing For the data presented here we used mean values for φ_{mean} and r_{mean} of all points in 8 9 PC_{plant}. Since we recorded time-lapse images of individual plants, we can also display 10 time-courses $\varphi_{\text{mean}}(t)$ and $r_{\text{mean}}(t)$ illustrated in Figure 4 as: 11 $\phi_{\text{mean}}(t) = \frac{1}{n_{\text{plant}}} \sum_{n=1}^{n_{\text{plant}}} \phi_{\text{mean}}(t, n)$ 12 Eq.1 $r_{\text{mean}}(t) = \frac{1}{n_{\text{plant}}} \sum_{n=1}^{n_{\text{plant}}} r_{\text{mean}}(t, n)$ 13 Eq.2 14 15 where t is the zeitgeber time and n_{plant} is the total number of measured plants. Hence values for φ_{mean} and r_{mean} measured for 15-30 individual plants were 16 17 averaged at a given time t. 18 We further aggregate $\varphi_{mean}(t)$ was used to compute: 19 1

20
$$\Delta\phi_{\text{mean}} = \frac{1}{n_{\text{plant}}} \left(\sum_{n=1}^{n_{\text{plant}}} \phi_{\text{mean}}(t_{\text{max}}, n) - \sum_{n=1}^{n_{\text{plant}}} \phi_{\text{mean}}(t_{\text{min}}, n) \right), \quad \text{Eq.3}$$

where $\Delta \varphi_{\text{mean}}$ is the daily amplitude of leaf movement (Fig. 4), t_{max} is the time during the 24h period, where leaves show the steepest elevation angle (around dusk) and t_{min} is the time during the 24h period, where leaves show the lowest elevation angle (usually 3h after dawn). Likewise, for $r_{\text{mean}}(t)$ we applied:

26

1
$$\Delta r_{\text{mean}} = \frac{1}{n_{\text{plant}}} \left(\sum_{n=1}^{n_{\text{plant}}} r_{\text{mean}} \left(27, n \right) - \sum_{n=1}^{n_{\text{plant}}} r_{\text{mean}} \left(3, n \right) \right), \qquad \text{Eq.4}$$

where Δr_{mean} represents a measure for the daily increase in rosette radius (Fig. 4). The non-parametric Wilcoxon Rank-Sum Test was used to test differences of measured traits between treatments, where small *P*-values indicate that independent samples follow continuous distributions with different medians.

6

7 **Results**

8 Diurnal pattern of ϕ_{mean} for the different treatments

9 In the following section we provide examples for leaf hyponasty in Arabidopsis for

10 different light conditions and genotypes using our phenotyping approach. The analysis

¹¹ of time-lapse images yielded φ_{mean} as a function of zeitgeber time *t* measured in hours

12 $\varphi_{\text{mean}}(t)$ (Eq. 3; see Fig. 5). Leaves moved up (increase in leaf elevation angle) during

the day and moved down during the night approaching a minimum elevation angle 2-3h

14 after dawn. An unexpected abrupt upward movement of leaves was observed under

15 high E_{PAR} and long day conditions, immediately after switching off the light (Fig. 5a).

16 This phenomenon was observed throughout all experiments under these conditions

17 (data not shown).

18 In short days, leaves showed a constitutively increased φ_{mean} compared to long day

19 conditions (Fig. 5a). Plants were already entrained in the respective photoperiod prior to

20 measurements and had already elevated leaves in short days upon transfer to the

21 phenotyping device. Surprisingly, $\varphi_{mean}(t)$ were virtually parallel for long and short

22 day, indicating that the overall diurnal pattern of up and downward leaf movement was

23 independent of day length (Fig. 5a). Moreover, the amplitude of leaf movement $\Delta \phi_{mean}$

24 was not different between both treatments in the two days (Fig. 5d, Column I,II,

25 *P*=0.022).

26 Next, we investigated how shade treatments using low E_{PAR} or high E_{FR} affected the

27 pattern of $\varphi_{mean}(t)$. Both led to increased values for φ_{mean} during the first hours (3 < *t* <

- 28 16; Fig. 5b), whereas the kinetics differed between the two treatments. Hyponastic
- 29 movement of leaves started right after the shade treatment had been initiated at t = 3.

1	The amplitude of leaf movement $\Delta \phi_{mean}$ was increased upon shade treatments
2	compared to the control during the first day (Fig. 5d; Column II-IV, P<0.001). After
3	the phase of downward movement (night and early morning), at $t = 27$ leaves did not
4	return to the initial elevation angle, which they had previously adopted at $t = 3$ the day
5	before. During the second day, $\Delta \phi_{\text{mean}}(t)$ followed the typical diurnal pattern, but values for
6	ϕ_{mean} remained constitutively increased (Fig. 5b). Compared to the control, estimated
7	values for $\Delta \phi_{\text{mean}}$ were bigger at high E_{FR} (Fig. 5d; Column II,III, P<0.001), but the
8	same for E_{PAR} (Fig. 5d; Column II, IV, $P=0.812$).
9	Lastly, we compared the response of the <i>sav3-2</i> mutant, affected in shade-induced (low
10	E_{PAR}) hyponasty, compared to Col-0 under low E_{PAR} (Fig. 5c). Initial values for φ_{mean} at
11	<i>t</i> =3 were the same magnitude for both genotypes. In the <i>sav3-2</i> mutant, φ_{mean} did not
12	increase to the same extent as for the Col-0 control after shade treatment using low
13	E_{PAR} (Fig. 5c). This decreased response is also expressed in smaller values for $\Delta \phi_{\text{mean}}$
14	in both days measured (Fig. 5d, Column IV-V, P<0.001). Thus, hyponastic response of
15	leaves to low E_{PAR} was impaired but not completely absent in the sav3-2 mutant.
16	
17	Diurnal pattern of r_{mean} for the different treatments
18	The second trait we have measured is r_{mean} for the different light conditions and
19	genotypes. The apparent increase in r_{mean} (Fig. 6) during the day coincides with the
20	increase in ϕ_{mean} and is a measurement artifact, which is discussed below. To compare
21	between treatments we only used values for r_{mean} are given for zeitgeber time $t=3, 27,$
22	51 (highlighted as open circles in Fig. 6), where leaves were at their most horizontal
23	position. These values were used to compute Δr_{mean} (Eq. 4).
24	Values for r_{mean} were smaller in short days compared to long days at the beginning of
25	measurements at $t=3$ (Fig. 6a) and the difference between the two treatments increased
26	towards t=51. Likewise, values for Δr_{mean} were smaller in short day compared to long
27	day (Fig. 6d, Column I-II, P<0.001), which in turn indicates a decreased leaf elongation
28	in response in short day compared to long day conditions.
29	Shade treatment using low E_{PAR} had no impact on the magnitude of r_{mean} (Fig. 6b).
30	Differences in Δr_{mean} were not significant (<i>P</i> >0.01). Shade treatment using high E_{FR} led

1 to increased values for r_{mean} at t=27 and t=51 compared to control (Fig. 6b). However,

2 Δr_{mean} during the first day was significantly increased compared to the control (Fig.

3 6d, Column II-III, *P*<0.001).

4 Furthermore, a clear phenotype regarding r_{mean} was observed for the *sav3-2* mutant,

5 which had smaller values for r_{mean} (Fig. 6c) and for Δr_{mean} compared to Col-0 control

6 (Column IV-V, P<0.001). This indicates that leaf elongation was impaired in the same

7 way as leaf hyponasty in the *sav3-2* mutant compared to Col-0 under the investigated

8 conditions. Our phenotyping approach is able to quantify this effect.

9

10 Discussion

11 We are interested in studying light-mediated growth responses in *Arabidopsis* and

12 understanding the underlying regulatory cellular and molecular processes. Previous

13 work has mainly focused on hypocotyl length as a phenotypic marker, but other traits

such as light-mediated leaf hyponasty or elongation have received less attention

15 (Millenaar et al. 2005; Mullen et al. 2006; Lorrain et al. 2008; Sasidharan et al. 2010;

16 Keller et al. 2011; Keuskamp et al. 2011; Polko et al. 2012). Our phenotyping method

17 based on laser scanning provides a non-invasive tool to measure the diurnal kinetics

18 of leaf hyponasty with high-throughput and high temporal resolution (<1h). Individual

19 plants can be subjected to two distinct fluence rates of PAR and different R/FR ratios

20 (Tab. 1). Entrained day-night cycles can be maintained, which allows monitoring of

21 plants during several days.

22 In this paper, we provide an initial characterization of our phenotyping device and the

23 developed image processing pipeline. The principal outputs of our method are time-

24 courses of two traits: i) $\varphi_{mean}(t)$ -a marker for leaf hyponasty- and ii) $r_{mean}(t)$ -a marker

25 for rosette radius or average leaf length. Both traits have been measured at a temporal

resolution of less than or equal to 1h during several days in *Arabidopsis* in an early

growth stage 1.05 (Boyes *et al.* 2001), where leaf length was usually smaller than 1.8

28 cm. This way 5-6 plants could be grown in one pot without overlapping of leaves

29 between plants. The maximum leaf length of plants is 3.9 cm to be entirely scanned.

30 In this case only two plants per pot can be scanned. The plant size limitation is clearly

31 a disadvantage of our approach, since care need to be taken that leaves of neighboring

32 plants do not overlap within the filter radius (here 1.8 cm) specified in the image

analysis. In case of overlapping measured points from neighboring plants perturb the
 measurements.

3 The geometric assembly of the laser scanner is optimized to measure horizontally 4 oriented leaves (Fig. 1a). However, leaf hyponasty leads to a situation, such that only 5 the upper part of leaves can be measured. In this case, certain parts of the plant are 6 masked by leaves and are therefore not visible to the camera or not lit by the laser. This 7 does not affect the correct assessment ϕ_{mean} , since leaves are relatively straight, but the 8 correct assessment of r_{mean} (Fig. 6) The apparent increase in r_{mean} -as leaves move up- is 9 caused by the incapacity of the laser scanner to record the lower parts of leaves (with 10 small values for r) at high leaf elevation angles. To compare between treatments, we therefore only used values r_{mean} 3h after dawn to compute Δr_{mean} (Eq. 4), when leaves 11 12 were oriented most horizontally, which was the optimal position for scanning. 13 Another potential application for our plant phenotyping method is the measurement of 14 architectural traits for individual leaves, such as petiole angle or leaf blade area. 15 Assuming the ideal case, where the complete leaf surface can be scanned and converted 16 into a 3D point cloud, automated surface triangulation can be applied as proposed by 17 Kaminuma et al. (2004). The resulting 3D surface could then be analyzed to directly 18 derive the traits of interest. In experimental conditions, leaves overlap or obscure 19 certain parts of the plant rendering the measured plant surface incomplete (see Fig. 4a 20 and 3D data in Supplemental material). To overcome this obstacle, a model based 21 reconstruction of individual leaf architecture could be applied as has been already 22 proposed for cereal leaves (Dornbusch et al. 2007). 23 Presented data on Arabidopsis leaf hyponasty in response to different shade treatments 24 agrees with data obtained in similar studies. High E_{FR} or low E_{PAR} leads to leaf 25 hyponasty starting shortly after the application of the shade treatment (Mullen et al. 26 2006; Millenaar et al. 2009; Sasidharan et al. 2010; van Zanten et al. 2010) and the 27 sav3-2 mutant is impaired in this shade-avoidance response (Tao et al. 2008; Moreno et 28 al. 2009; Keller et al. 2011). However, these studies usually have measured the leaf 29 responses to shade during one day and constant light, or at a single time point. 30 According to our data, the diurnal magnitude of up- and down leaf movement follows 31 the pattern of the control in the second day after the shade treatment, except that leaves 32 remain constitutively elevated (Fig. 5).

1 Mullen *et al.* (2006) made measurements of the diurnal pattern of leaf hyponasty in 12h 2 day-night cycles, which looks very similar to our data. They also show an abrupt 3 upward movement of leaves immediately after switching off the light (Fig. 5a). 4 However the underlying mechanisms have not been discussed. A mutant screen could 5 help to identify genes involved in this growth response. It seems that long days ($\geq 12h$) 6 and high E_{PAR} promote this effect, whereas short day and/or low E_{PAR} inhibit it. 7 To our knowledge, hyponastic movements of single leaves have not been measured 8 following several days. Thus it is not known how long leaves do follow diurnal up- and 9 down movements and whether cell elongation in the petiole is a prerequisite. These 10 questions could be addressed with our phenotyping method. 11 Wiese et al. (2007) have measured the diurnal pattern of leaf growth in Arabidopsis in 12 12h day-night cycles. They show a peak growth rate shortly after dawn, which coincides 13 with the phase of rapid downward movement of leaves in our data (Fig. 5). There is thus 14 a correlation between the rapid downward movement of leaves and increased leaf 15 growth. This leads to the unanswered question of whether the regulation of leaf 16 hyponasty and elongation rely on the same genetic regulatory network. On the one 17 hand, leaf hyponasty is mainly driven by differential cell elongation between the abaxial 18 and adaxial epidermis cells at the very base of the petiole (Polko *et al.* 2012). On the 19 other hand, petiole growth by cell proliferation and elongation is biggest towards the 20 blade-petiole junction (Ichihashi et al. 2011). Thus both growth processes are spatially 21 separated. One may further speculate that the regulatory mechanisms leading to leaf 22 hyponasty and elongation are somewhat different because of this spatial separation. We 23 would like to address this issue in future work and apply the phenotyping method that 24 has been presented in this paper.

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Tab. 1: The four light setups (I-IV) adjustable in our phenotyping device characterized by mean values (standard deviation) of the fluence rate of photosynthetically active radiation (E_{PAR}) or far-red radiation (E_{FR}) for each setup Arrows indicate high (\uparrow or low ($\downarrow E_{\text{PAR}}$ and E_{FR} , respectively. Red color highlights deviation of light setup from control conditions.

Light setup	$E_{\rm PAR}$	E _{FR}
	$(\mu mol m^{-2} s^{-1})$	$(\mu mol m^{-2} s^{-1})$
I. $\uparrow E_{\text{PAR}} \downarrow E_{\text{FR}}$	150 (10)	4.5 (0.3)
II. $\uparrow E_{\text{PAR}} \uparrow E_{\text{FR}}$	150 (10)	7.5 (0.3) to 150 (10)
III. $\downarrow E_{\text{PAR}} \downarrow E_{\text{FR}}$	35 (3)	1.5 (0.2)
IV. $\downarrow E_{\text{PAR}} \uparrow E_{\text{FR}}$	35 (3)	2.5 (0.3) to 50 (4)



Fig. 1: (a) Laser scanner and charged coupled device (CCD) camera mounted on the imaging unit in the Scanalyzer HTS phenotyping device; (b) 2.5D height scaled image (1024-by-1728 pixels) containing six Arabidopsis plants (image provided as Supplemental Material). For a better visualization, the color in (b) was adjusted compared to the recorded original image. (c) 3D point cloud computed from the recorded 2.5D image using our developed image conversion. The Matlab script and required data for image conversion are available in the Supplemental material. 320x548mm (72 x 72 DPI)



Fig. 2: (a) Panel with 23 circles scanned with the ScanAlyzer HTS at an angle of 20°, (b) point cloud representing these circles in 3D after image conversion, (c) circle area computed from measured 3D points Ascan vs. predefined circle area Aorg on the panel. The dashed line is the 1:1 line. Values for the coefficient of determination (R2) and mean relative error (MRE) are given. 332x642mm (72 x 72 DPI)



Fig. 3: (a) Measured 3D point cloud of an Arabipopsis plant (PCplant); P0 = selected basal plant point, which coincides with the position of the shoot apical meristem of that plant; Pn = sample point to illustrate the definition of its spherical coordinates (θ n, φ n, rn); (b-d) Histograms illustrating the distribution of θ , φ and r for points in PCplant given in (a): b) θ vs. n, c) φ vs. n, d) r vs. n; where n = number of points per bin. 305x661mm (72 x 72 DPI)



Fig. 4: Diurnal pattern of the φ mean(t) -a marker for mean elevation angle- as solid line and rmean(t) – a marker for mean rosette radius- as dashed line. The gray box indicates the entrained night period, which was maintained during measurements. $\Delta \varphi$ mean as the measure for the daily amplitude of leaf movement; Δr mean as the measure for the daily increase in rosette radius. 357x204mm (72 x 72 DPI)



Fig. 5: (a-c) Diurnal pattern of opmean as a marker the mean elevation angle of plants. Dots represent mean values for omean computed from 15-30 individual plants. The gravish band around mean values for opmean represent the 95% confidence interval of mean estimate. Vertical gray boxes indicate the entrained night period in long day and the hatched part the supplemental night time in short day conditions. Plants were grown in a separate incubator at 21°C, EPAR=180 µmol m-2 s-1 and photoperiod for 15 days (long day) or 20 days (short day) prior to measurements. Specific treatments: (a) black line: long day; red line: short day; for both treatments: Col-0 at normal EPAR=150 µmol m-2 s-1 and normal EFR =4.5 µmol m-2 s-1; (b) black line: normal EPAR=150 µmol m-2 s-1, EFR =4.5 µmol m-2 s-1; red line: normal EPAR=150 µmol m-2 s-1 and supplemental EFR =22.5 µmol m-2 s-1; blue line: reduced EPAR=39 µmol m-2 s-1; reduced EFR =1.5 µmol m-2 s-1); (c) solid line: Col-0; dashed line: sav3-2 mutant; reduced EPAR=39 µmol m-2 s-1; EFR =1.5 μ mol m-2 s-1; (d) boxplots of the daily amplitude of leaf movement $\Delta \phi$ mean measured for two consecutive days for different treatments presented in (a-c); from left to right: (I) Col-0, EPAR=150 µmol m-2, EFR =4.5 µmol m-2 s-1, short day; (II) Col-0, EPAR=150 µmol m-2, EFR =4.5 µmol m-2 s-1, long day, (III) Col-0, EPAR=150 µmol m-2, EFR =22.5 µmol m-2 s-1, long day; (IV) Col-0, EPAR=39 µmol m-2 s-1; EFR =1.5 µmol m-2 s-1, long day, (V) sav3-2 mutant; EPAR=39 μ mol m-2 s-1; EFR =1.5 μ mol m-2 s-1, long day. 724x539mm (72 x 72 DPI)



Fig. 6: (a-c) Diurnal pattern of rmean as a marker for the mean rosette radius or average leaf length of plants. Dots represent mean values for rmean computed from 15-30 individual plants. The grayish band around mean values for rmean represent the 95% confidence interval of mean estimate. Vertical gray boxes indicate the entrained night period in long day and the hatched part the supplemental night time in short day conditions. Plants were grown in a separate incubator at 21°C, EPAR=180 µmol m-2 s-1 and photoperiod for 15 days (long day) or 20 days (short day) prior to measurements. Specific treatments: (a) black: long day; red: short day; for both treatments: Col-0 at normal EPAR=150 μ mol m-2 s-1 and normal EFR =4.5 μ mol m-2 s-1; (b) black: normal EPAR=150 µmol m-2 s-1, EFR =4.5 µmol m-2 s-1; red: normal EPAR=150 µmol m-2 s-1 and supplemental EFR =22.5 µmol m-2 s-1; blue: reduced EPAR=39 µmol m-2 s-1; reduced EFR =1.5 µmol m-2 s-1); (c) solid line: Col-0; dashed line: sav3-2 mutant; reduced EPAR=39 µmol m-2 s-1; EFR =1.5 μ mol m-2 s-1. (d) boxplots of the daily increase in rosette radius Δ rmean measured for two consecutive days for different treatments presented in (a-c); from left to right: (I) Col-0, EPAR=150 µmol m-2, EFR =4.5 µmol m-2 s-1, short day; (II) Col-0, EPAR=150 µmol m-2, EFR =4.5 μmol m-2 s-1, long day, (III) Col-0, EPAR=150 μmol m-2, EFR =22.5 μmol m-2 s-1, long day; (IV) Col-0, EPAR=39 µmol m-2 s-1; EFR =1.5 µmol m-2 s-1, long day, (V) sav3-2 mutant; EPAR=39 μ mol m-2 s-1; EFR =1.5 μ mol m-2 s-1, long day. 726x538mm (72 x 72 DPI)