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Light Potentials of Photosynthetic Energy Storage in the Field: What limits the ability to use or dissipate rapidly increased light energy?

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Complete List of Authors:	Kanazawa, Atsuko; Michigan State University, MSU-DOE Plant Research Lab Chattopadhyay, Abhijnan; Michigan State University, MSU-DOE Plant Research Lab; Michigan State University, Department of Statistics Kuhlgert, Sebastian; Michigan State University, MSU-DOE Plant Research Lab Tuitupou, Hainite; Michigan State University, MSU-DOE Plant Research Lab Maiti, Tapabrata; Michigan State University, Department of Statistics Kramer, David; Michigan State University, MSU-DOE Plant Research Lab; Michigan State University, Biochemistry and Molecular Biology
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Author-supplied statements

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Ethics

Does your article include research that required ethical approval or permits?:

This article does not present research with ethical considerations

Statement (if applicable):

CUST_IF_YES_ETHICS :No data available.

Data

It is a condition of publication that data, code and materials supporting your paper are made publicly available. Does your paper present new data?:

Yes

Statement (if applicable):

Primary data is available on the photosynq.org site under the project "rapid-ps-responses-pam-ecst-npqt-mint-dmk". Data cleaning and analysis code is available in a GitHub repository (<https://github.com/protonzilla/Light-Potentials-in-Field>).

Conflict of interest

I/We declare a competing interest

Statement (if applicable):

D.M.K. and S.K. are co-founders of PhotosynQ which maintains the PhotosynQ platforms and distributes and maintains the MultispeQ instruments. The current project was performed independently with no funding to or from the PhotosynQ organization.

Authors' contributions

This paper has multiple authors and our individual contributions were as below

Statement (if applicable):

A.K. and D.M.K. designed the experiments. A.K. and H.T. conducted experiments. A.K., A.C., S.K. and D.M.K. analyzed data. A.K., A.C., S.K., T.M. and D.M.K. contributed to the interpretations of data and writing manuscript.

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3 **Light Potentials of Photosynthetic Energy Storage in the Field: What limits**
4 **the ability to use or dissipate rapidly increased light energy?**
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10 **Atsuko Kanazawa,^{1,2} Abhijnan Chattopadhyay^{1,3}, Sebastian Kuhlger¹, Hainite**
11 **Tuitupou¹, Tapabrata Maiti³ and David M. Kramer^{1,4*}**
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15
16 ¹ MSU-DOE Plant Research Lab, Michigan State University, East Lansing, MI 48824,
17 USA

18
19 ² Chemistry, Michigan State University, East Lansing, MI 48824, USA

20
21 ³ Department of Statistics and Probability, Michigan State University, East Lansing, MI
22 48824, USA

23
24 ⁴ Biochemistry and Molecular Biology, Michigan State University, East Lansing, MI
25 48824, USA
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30 * Correspondence: kramerd8@msu.edu
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32 **Summary.** The responses of plant photosynthesis to rapid fluctuations in environmental
33 conditions are thought to be critical for efficient capture of light energy. Such responses are
34 not well represented under laboratory conditions, but have also been difficult to probe in
35 complex field environments. We demonstrate an open science approach to this problem that
36 combines multifaceted measurements of photosynthesis and environmental conditions, and
37 an unsupervised statistical clustering approach. In a selected set of data on mint (*Mentha*
38 sp.), we show that the “light potential” for increasing linear electron flow (LEF) and
39 nonphotochemical quenching (NPQ) upon rapid light increases are strongly suppressed in
40 leaves previously exposed to low ambient PAR or low leaf temperatures, factors that can act
41 both independently and cooperatively. Further analyses allowed us to test specific
42 mechanisms. With decreasing leaf temperature or PAR, limitations to photosynthesis during
43 high light fluctuations shifted from rapidly-induced NPQ to photosynthetic control (PCON)
44 of electron flow at the cytochrome *b₆f* complex. At low temperatures, high light induced
45 lumen acidification, but did not induce NPQ, leading to accumulation of reduced electron
46 transfer intermediates, a situation likely to induce photodamage, and represents a potential
47 target for improving the efficiency and robustness of photosynthesis. Finally, we discuss the
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3 implications of the approach for open science efforts to understand and improve crop
4 productivity.
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9 **Keywords:** nonphotochemical quenching (NPQ), photodamage, proton motive force, qE,
10 photosynthetic control, unsupervised learning
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17 **Introduction**

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19 While oxygenic photosynthesis supplies energy to drive essentially all biology in our
20 ecosystem, it involves highly energetic intermediates that can generate highly toxic reactive
21 oxygen species (ROS) that can damage the organisms it powers [1]. Thus, the energy input
22 into photosynthesis must be tightly regulated by photoprotective mechanisms that act at
23 several key steps in the light reactions. The balance and kinetics of this regulation is an
24 active target for crop improvement.
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31 One class of photoprotective processes, known as nonphotochemical quenching
32 (NPQ), dissipates absorbed light energy as heat, thus diverting energy away from
33 photosystem II (PSII) [2], decreasing the accumulation of reactive intermediates. This
34 photoprotective capacity comes at the cost of decreased photochemical efficiency, and thus
35 the organisms must regulate NPQ to balance the avoidance of photodamage with efficient
36 energy conversion [3,4]. There are several forms of NPQ that differ in their mechanisms
37 and rates of activation and deactivation. The most rapid NPQ form is qE, which is activated
38 by acidification of the thylakoid lumen by the proton gradient (ΔpH) component of the
39 thylakoid proton motive force (*pmf*) [2]. Lumen acidification activates the violaxanthin
40 deepoxidase or VDE [5–8] resulting in the conversion of violaxanthin (Vx) to
41 antheraxanthin (Ax) and zeaxanthin (Zx); and protonation of PsbS, an antenna-associated
42 protein required for qE [2], which appear to act cooperatively in setting the extent of qE.
43 The conversion of Vx to Ax and to Zx is typically much slower than the rapidly reversible
44 protonation of PsbS [2], and during prolonged illumination, the responses of qE will likely
45 be limited by the rate of acidification and de-acidification of the thylakoid lumen, which
46 are, in turn, governed by ion movements in the chloroplasts [9–11]. Slower forms of NPQ
47 have also been demonstrated [12], including qI, which is related to the photodamage and
48 repair of photosystem II (PSII) or qZ, which related to the accumulation of Zx
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3 (independently from qE) [13], qH, related to cold and high light stress [13], and qT, related
4 to antenna state transitions [14].
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8 The acidification of the thylakoid lumen also controls electron transfer at the cytochrome
9 *b₆f* complex, a process called photosynthetic control (PCON) [15–20], which prevents the
10 buildup of electrons on the acceptor side of photosystem I (PSI) that can lead to
11 photodamage [15,21–23]. Interestingly, PCON and qE (both responses to lumen
12 acidification) are expected to have opposing effects on Q_A redox state. High levels of
13 PCON in the absence of qE would lead to accumulation of plastoquinol (PQH₂) and the
14 reduced form of the PSII electron acceptor, Q_A⁻, which can potentiate photodamage. Thus,
15 these two processes must be tightly coordinated, with qE being activated at lumen pH
16 somewhat less acidic than PCON [15].
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24 Plants in natural environments are exposed to rapidly changing environmental conditions,
25 especially light which can change by orders of magnitude in less than a second. It has
26 become clear that rapid and unpredictable fluctuations in light intensity can be more
27 damaging than more gradual changes [22,24–31]. This sensitivity can partly be related to
28 the buildup of reactive redox intermediates and thylakoid *pmf*, which can occur following
29 low-to-high light transitions much more rapidly than the activation of photoprotective NPQ
30 and PCON, leaving the photosynthetic apparatus prone to photodamage. Also, the slow
31 recovery of NPQ following a decrease in light intensity can lead to substantial losses of
32 photosynthetic efficiency [32]. Recently, it has been reported that engineering plants with
33 increased expression levels of VDE and zeaxanthin epoxidase (ZE), resulted in accelerated
34 formation and reversal of qE accompanied by increased plant productivity [3], suggesting
35 that it may be possible to increase yield in crops by modifying photosynthetic regulatory
36 responses.
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48 On the other hand, we lack comprehensive surveys of the range of natural response of
49 photosynthesis to real environmental fluctuations, in part because of a lack of deployable
50 scientific equipment and methods to probe these processes in the field. Consequently, it has
51 not been possible to assess the mechanistic bases of extant natural variations in these
52 processes, their possible benefits or tradeoffs, or which of these may be most useful for crop
53 improvement.
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3 Here, we introduce a method and proof-of-concept field data results to address the
4 following questions: Can we assess the extent of natural variations in rapid responses to
5 fluctuations in photosynthetically-active radiation (PAR) intensity for both electron flow
6 and photoprotection? How do these limitations depend on environmental conditions? What
7 are the mechanisms that underlie these variations in responses to rapidly fluctuating light in
8 the field?
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12 Here, we introduce an approach to both measure and analyse these variations, focusing on
13 one species, *Mentha* sp., under a limited set of conditions, and applied these to testing
14 among a set of mechanisms that can be distinguished based on a range of optical
15 measurements available using the MultispeQ 2.0 device, including: 1) PSI acceptor-side
16 limitations to electron transfer; 2) Increased NPQ which limits the input of light energy into
17 photosystem II (PSII); and 3) Photosynthetic control in which acidification of the lumen
18 slows electron transfer at the level of plastoquinol (PQH₂) oxidation by the cytochrome *b₆f*
19 complex.
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22 The results show that the approach can effectively be used to assess the range of variations
23 in ‘light potentials,’ the extent to which increased light leads to increased photosynthetic
24 responses, under field conditions, as well as to test specific hypothetical models, setting up
25 a broad-scale, multiple participant, open science approach to exploring the responses across
26 multiple species, genotypes and environments. The results also reveal, at least in *Mentha*,
27 unexpected leaf temperature-dependent limitations in the rapid formation of NPQ that result
28 in the accumulation of reduced PSII electron acceptor, Q_A and a high thylakoid *pmf*,
29 conditions likely to promote the formation of reactive oxygen species.
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45 **Materials and Methods**

46 **Plants and leaf sampling.**

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49 Measurements were made in a population of *Mentha* sp. (mint) plants that have been
50 maintained at the MSU Horticulture Gardens for at least 10 years. The GPS locations of all
51 measurements are included in the online data set
52 (<https://photosynq.org/projects/rapid-ps-responses-pam-ecst-npqt-mint-dmk>). Although it
53 was not practical to exhaustively capture the lifecycle of the plants, the experimental
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3 strategy sampled a sufficiently wide range of conditions to allow clear patterns emerged in
4 the relationships between phenotypes and environmental parameters, as described below.
5 The experiment took place over a nine-day experimental window (Figure S1A), sampling a
6 range of times of day, temperatures etc. (Fig. S1B). Measurements were made at multiple,
7 alternating canopy levels and positions (subjectively at high, middle and low canopy levels)
8 from early morning, though later afternoon (Fig. S1B), and at multiple locations across the
9 plots on each day.
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17 **Measurements of photosynthetic and related parameters using MultispeQ 2.0.**

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21 Optical measurements were made using MultispeQ V2.0 hand-held instruments
22 (<https://photosynq.com>), based on that presented by Kuhlger et al [33] and calibrated using
23 the CaliQ calibration system (<https://photosynq.com/caliqu>). The Light Potential (LP)
24 protocol used in the experiments can be found in the online project information
25 ([rapid-ps-responses-with-ecs-fast-ecs-dirk-and-npqt-dmk](https://photosynq.com/rapid-ps-responses-with-ecs-fast-ecs-dirk-and-npqt-dmk)) as illustrated in Figure 1. The
26 protocol was designed to strike a balance between needs for sampling large numbers of
27 leaves, the desire for detailed spectroscopic measurements and the length of time the plant
28 could be exposed to increased or decreased PAR. The full protocol, with measurements at
29 ambient, after 10 s full sunlight and 10 s dark required about 35-40 s, at the limit of the time
30 scale over which most researchers could steadily clamp a leaf in the instrument. The
31 implications of the 10 s illumination and recovery time are discussed in the Results and
32 Discussion sections.
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43 In the first stage of the protocol (Fig. 1A), the MultispeQ was programmed to continuously
44 (at about 5 Hz) measure photosynthetically-active radiation (PAR) and reproduced these
45 levels using a red actinic LED (655 nm emission peak) illuminating the adaxial surface of
46 the leaf. When the MultispeQ detected that a leaf was clamped in the chamber, a series of
47 measurement sequences were initiated. After a few seconds of illumination at ambient PAR
48 (PAR_{amb}) to allow for settling and setting of gains, the first set of measurements was made,
49 estimating at PAR_{amb} LEF (LEF_{amb}), NPQt (NPQ_{amb}) and other photosynthetic parameters
50 (Fig. 1B).
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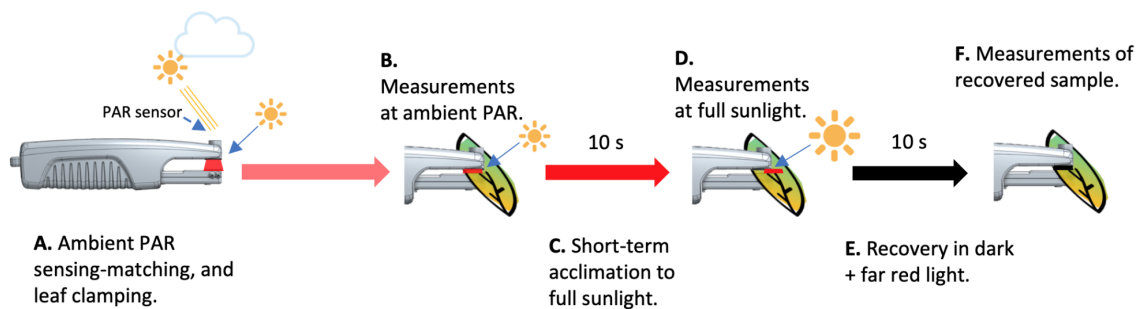


Figure 1. Experimental procedure of NPQ Light Potential designed to detect the change in NPQ induced under different light intensities. A) A sensor on the MultispeQ continuously detects the ambient photosynthetically active radiation (PAR) in the field and reproduces this PAR value using internal LED. B) When the leaf clamp is closed over a leaf, the experiment begins by recording the local PAR, leaf temperature, ambient temperature, leaf angle and GPS position. After a short period of illumination at the measured ambient PAR, the first set of optical measurement are recorded. C) Once completed, the leaf is exposed to a period of high PAR ($2,000 \mu\text{mol}\cdot\text{m}^{-2}\cdot\text{s}^{-1}$, equivalent to full sunlight) for 10 s. D) The optical measurements are repeated at high PAR. E) The leaf is then dark adapted (actinic light is switched off), with weak far red background light for 10 s (Panel E). F) A final set of optical measurements is made to assess rapid dissipation of NPQ and reoxidation of accumulated reduced intermediates. Each set of optical measurements includes chlorophyll fluorescence and absorbance changes to give estimates of ϕ_{II} , LEF, NPQt, q_L (Q_A redox state); ECSt, P_{700} redox state, and g_{H^+} (relative proton conductivity of the thylakoid ATP synthase), as described in Materials and Methods. Measurements taken at ambient and light are designated as “measurement (ambient), LEF (ambient), NPQt (ambient), q_L (ambient).

The actinic light was then increased to approximately full sunlight ($2,000 \mu\text{mol}\cdot\text{m}^{-2}\cdot\text{s}^{-1}$ red light) for 10 s (Fig. 1 C), after which the photosynthetic measurements were repeated (Fig 1D), yielding measurements of LEF_{high} , NPQ_{high} etc. We chose full sunlight, rather than an artificially intense super-saturation light, to estimate light potentials that could occur in the field, and not the absolute maximum, and to avoid non-physiological or photoinhibitory effects. Thus, the light potentials of various processes will be limited as PAR_{amb} approaches full sunlight.

In the third stage of the experiment, the actinic light was then switched off, and a weak far-red light switched on for 10s (Fig. 1E), following another repetition of the measurements to assess the extent of NPQ_t after relaxation (NPQ_{rec} , Fig. 1F).

Environmental parameters including PAR, temperature, humidity, leaf temperature, leaf

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3 angle and GPS location were measured either prior to or following the physiological
4 measurements.
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8 Chlorophyll fluorescence changes were measured using MultispeQ 2.0 devices to estimate
9 PSII quantum efficiency (Φ_{II}) and linear electron flow (LEF) [34,35], as well as qL, a
10 measure of the fraction of Q_A in the oxidised state [36], and the extent of NPQ based on the
11 rapid “total” NPQ method developed by Tietz et al. [37], designated as NPQ_t. Just prior to
12 the saturation pulses, dark interval relaxation kinetics (DIRK, dark interval of
13 approximately 300 ms) of the absorbance changes around 520 nm attributed to the
14 electrochromic shift (ECS) were recorded. Fitting the ECS signals to exponential decay
15 curves yielded estimates of the relative light-dark differences in thylakoid *pmf* (ECS_t) and
16 the proton conductivity of the chloroplast ATP synthase (g_{H^+}), as described in [16,38,39].
17 To account for differences in leaf thickness, light path or number of chloroplasts in various
18 leaves, the ECS_t values were normalised to the relative chlorophyll contents as estimated by
19 the SPAD parameter [33], which was measured at the end of the experiment. The extent of
20 oxidation of P₇₀₀ in the light was estimated by the DIRK of infrared LED light using an
21 LED measuring pulse with peak emission at ~810 nm.
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34 **Environmental conditions during light potential measurements in the field.**

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38 Supplemental figures S1A-C show the distributions of environmental factors (light
39 intensities, leaf temperatures) for the measurements analyzed in this study. The MultispeQ
40 sensor was positioned by the user to be parallel to the leaf surface, so that the
41 cosine-corrected PAR sensor should effectively estimate PAR absorbed by the leaf surfaces
42 *in situ* throughout their canopy, and thus the ambient PAR (PAR_{amb}) values were dependent
43 on both time of day (diurnal cycle, Fig. S1B) and by leaf angle (Fig. S1C). Ambient
44 temperature and leaf temperatures (T_{leaf}) were dependent on time of day, with obvious
45 influences from weather-related fluctuations (Fig. S1A, B). We chose to compare results to
46 T_{leaf}, rather than ambient temperature, to better reflect the effects on leaf photosynthetic
47 processes.
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55 **Data calculations and cleaning**

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3 Data from the PhotosynQ platforms was reprocessed and cleaned to improve the estimation
4 of decay constants for electrochromic shift and near infrared absorbance changes. As with
5 any field experiments, some results were found to have obvious errors or be out of
6 acceptable ranges, and were removed from the analysis. However, all original data was
7 maintained in the online platform, allowing the reader to explore and reanalyse the effects
8 of our data cleaning procedures. The rules and code for data flagging are defined in the
9 Jupyter Notebook, (see Supplemental Information, “Data Cleaning Notebook”). A total of
10 292 points were flagged from a total of 1346 original measurements. The majority of the
11 flagged measurements (179) were due to a defective device. The remaining 113 flagged
12 points can be attributed to user error (e.g. leaf movements during measurements) or poor
13 signal-to-noise that resulted in parameter values outside the theoretical ranges.
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24 **Results**

27 **Field measurements of photosynthetic parameters under ambient and rapidly elevated** 28 **PAR.**

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33 Figure 2A shows LEF measured at PAR_{amb} (LEF_{amb}) plotted against ambient PAR_{amb} and
34 leaf temperature (T_{leaf} , see colouration of points). The plots use the square root of PAR to
35 better resolve the results at lower PAR_{amb} , and to partially linearise the responses. LEF_{amb}
36 increased with increasing PAR_{amb} , with a roughly hyperbolic dependence and an apparent
37 half-saturation point of about $350 \mu\text{mol photons m}^{-2} \text{s}^{-1}$, reaching maximum values of about
38 $250 \mu\text{mol electrons m}^{-2} \text{s}^{-1}$ at $1700 \mu\text{mol photons m}^{-2} \text{s}^{-1}$.
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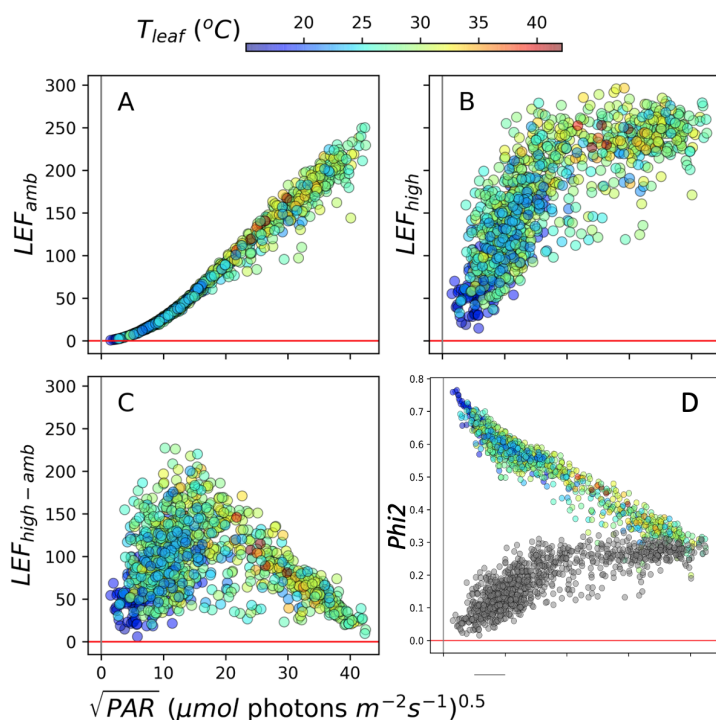


Figure 2. Light and temperature effects on linear electron flow (LEF) and Photosystem II quantum efficiency (Φ_{II}). Each parameter was plotted functions of the square root of the ambient photosynthetically active radiation (PAR_{amb} , X-axis) and Leaf Temperature (T_{leaf} , coloration of points). A) Dependencies of Linear Electron Flow (LEF) measured at PAR_{amb} ; B) LEF measured at 10s high light (LEF_{high}); C) The high light-induced differences in LEF ($LEF_{high-amb}$); D) The PSII quantum efficiencies measured under ambient PAR (Φ_{IIamb} , points colored by T_{leaf}) and at 10s high light (Φ_{IIhigh} , grey points).

Upon ten seconds of exposure to $2000 \mu\text{mol photons m}^{-2} \text{s}^{-1}$ increased LEF to generally higher values (LEF_{high} Fig. 2B), indicating that LEF_{amb} was at least partly light-limited under all of the conditions. Note that each LEF_{amb} point was taken on different leaves at different times (Materials and Methods) and has corresponding LEF_{high} and $LEF_{high-amb}$ measurements. The relationship between measurements is illustrated in Fig. S2, which shows selected pairs of LEF_{amb} and LEF_{high} connected by vertical line segments. The extent of LEF_{high} was not uniform, but appeared to be strongly suppressed at low PAR_{amb} and/or low T_{leaf} . The high light-induced difference in LEF ($LEF_{high-amb}$) increased with PAR_{amb} at low light, reaching a peak at about $200 \mu\text{mol photons m}^{-2} \text{s}^{-1}$, above which it declined as PAR_{amb} approached PAR_{high} and LEF_{high} became light-saturated. The suppression of LEF_{high} was due to large decreases in the quantum efficiencies of PSII (Φ_{II} , Fig. 2D). Φ_{II} at PAR_{amb} (Φ_{IIamb}) were highest at low PAR_{amb} , and progressively saturated as light was increased. The opposite behavior was seen with Φ_{II} measured after 10s of high light

($\Phi_{2\text{high}}$ Fig. 2D, grey symbols) which was lowest at low PAR_{amb} , and gradually increased with PAR_{amb} .

Gaussian Mixture Model clustering analysis of field data

A simple linear effects model applied over the entire data set (Supplemental Table S1A) indicated strong correlations between LEF_{amb} and both PAR_{amb} and T_{leaf} , suggesting that both environmental factors controlled LEF_{amb} . However, such correlations may be coincidental since PAR and T_{leaf} are both expected to be dependent on weather or time of day, as it is clear from the strong statistical correlations between PAR and T_{leaf} . Also, the effects are likely to be co-dependent. For example, at low PAR_{amb} , LEF_{amb} should be light-limited, and thus have minimal dependence on T_{leaf} , but at higher PAR_{amb} , may be more strongly controlled by temperature-dependent processes.

One approach to disentangling these effects would be to slice the data into segments, e.g., at different ranges of PAR_{amb} , and test for correlations with T_{leaf} within each segment. However, arbitrary-chosen ranges for the segments can add bias, or fail to detect more complex interactions. We thus applied a Gaussian Mixture Model (GMM) clustering approach based on those presented earlier [40,41]. Because GMM is an unsupervised machine learning method, it can reduce bias in the selection of clusters that represent regions of distinct interactions among environmental and photosynthetic parameters. GMM assumes that the data points from the population of interest are being drawn from a combination (or mixture) of Gaussian distributions with certain parameters, and performs an optimization scheme to a sum of K Gaussian distributions, allowing for a completely unsupervised process, avoiding potential user bias. An expectation-maximization (EM) algorithm was used to fit the GMM to the dataset, generating a series of Gaussian components (clusters) with distributions characterised by specific means and covariance matrices. The optimal number of clusters was determined using the Bayesian Information Criterion (BIC), the value of the maximised log likelihood, with a penalty on the number of parameters in the model [40–43]. This approach also allows comparison of models with differing parameterizations and/or differing numbers of clusters, because the volumes, shapes, and orientations of the covariances can be constrained to those described by defined models [40].

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3 Clusters obtained through GMM have both within cluster (intracluster) and between cluster
4 (intercluster) variations. In order to test for intercluster variation, we used the clustering
5 assignment obtained for one phenotype and applied it on other phenotype(s). Here we want
6 to investigate what would be the distinctive behaviour of different phenotypes if we have
7 used the same configuration. Using the same set of cluster assignments to different
8 phenotypes, one might be skeptical of the clustering behaviour as phenotypes interact
9 differently with PAR_{amb} and T_{leaf} . In that case, we might not be able to directly compare the
10 inter cluster behaviours of phenotypes. To mitigate this issue, we use the GMM clustering
11 as a tool to create a “baseline” clustering configuration for one phenotype and use that
12 configuration over another phenotype. We set up our hypothesis as two phenotypes are
13 similar under the same configuration against they are not. If the interaction pattern of one
14 phenotype with PAR_{amb} and T_{leaf} changes over the other phenotype, we reject our hypothesis
15 and imply that different configurations of PAR_{amb} and T_{leaf} interact differently with
16 phenotypes. By doing this we are able to disentangle the effect of PAR_{amb} and T_{leaf} and infer
17 regarding the intracluster variations as to be a key element to determine variations in the
18 interactions between parameters and variations in environmental conditions, e.g., to assess
19 if a relationship is modulated in different ways under different ranges of conditions. Also,
20 as will be seen in the Discussion, intercluster variations (differences in the mean and
21 covariances between clusters) can be used to differentiate distinct patterns of behavior, or
22 mechanistic interactions, between conditions.
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39 As shown in Fig. S3, GMM analysis of LEF_{amb} , PAR_{amb} and T_{leaf} , found six distinct, compact
40 clusters that differed in the mode of interaction among the photosynthetic and
41 environmental parameters. Encompassing points with lower PAR_{amb} showed strong
42 (Clusters 1,2,4 and 5) dependence of LEF_{amb} on PAR_{amb} , with little contributions from T_{leaf} .
43 By contrast, two clusters (3 and 6), which included points at higher PAR_{amb} , showed
44 substantial dependencies on both PAR_{amb} and T_{leaf} . These results are consistent with LEF
45 being predominantly light-limited at low ambient PAR, but increasingly limited by
46 temperature-dependent processes at higher PAR. The presence of these two classes of
47 clusters indicates that PAR_{amb} and T_{leaf} are likely to affect LEF_{amb} in independent ways. The
48 fact that the shapes of the clusters were not determined with individual slicing under the
49 individual parameters for PAR_{amb} and T_{leaf} , but with a co-dependence on both PAR_{amb} and
50 T_{leaf} , suggests that, under some conditions, these effects interact, e.g. T_{leaf} may affect the
51 dependence of LEF_{amb} on PAR_{amb} .
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GMM identified five distinct clusters for interactions among LEF_{high} , PAR_{amb} and T_{leaf} (Fig. S4). In contrast to the results on LEF_{amb} , clusters at lower PAR_{amb} (1, 2 and 4) showed LEF_{high} dependencies on both T_{leaf} and PAR_{amb} , while Cluster 3 showed correlations with T_{leaf} , but not with PAR_{amb} . The stronger dependence on T_{leaf} of LEF_{high} compared to LEF_{amb} implies that the exposure to high light revealed additional rate limitations in LEF_{high} that were more strongly controlled by both T_{leaf} and PAR_{amb} and that, at least under some conditions, these effects were independent of each other.

Analysis of NPQ

NPQ_t measured under PAR_{amb} (NPQ_{amb} , Fig. 3A) showed a positive correlation to PAR_{amb} , with an apparent tendency for smaller values at lower T_{leaf} . NPQ_{amb} showed considerable variations, compared to LEF_{amb} , even at low PAR_{amb} , consistent with the idea that NPQ is governed not only by PAR but by metabolic, developmental or other environmental parameters.

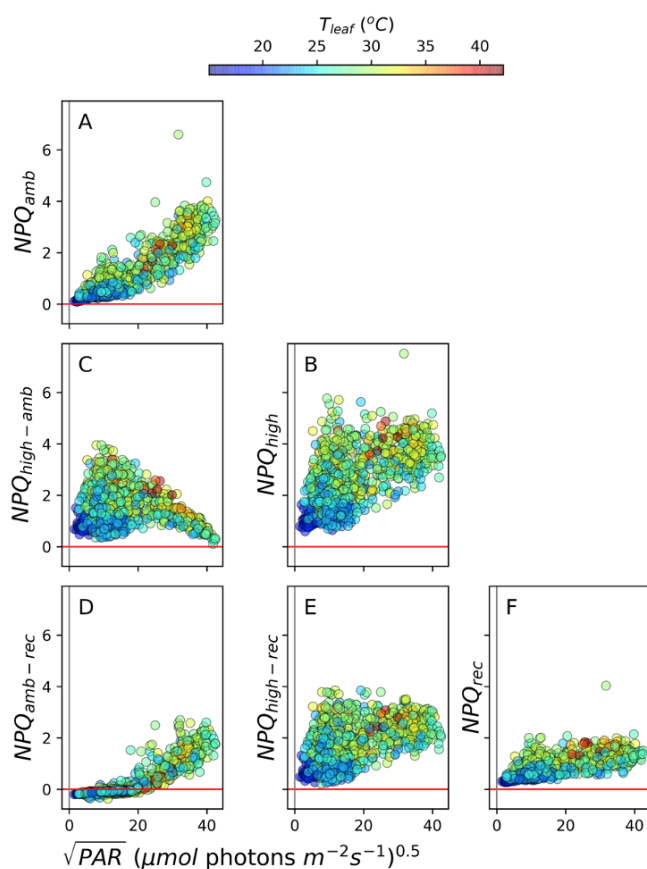


Figure 3. Light and temperature effects on NPQ. The NPQ parameter was plotted functions of the square root of the ambient photosynthetically active radiation (PAR_{amb} , X-axis) and Leaf Temperature (T_{leaf} , coloration of points). A) Induced NPQ measured at PAR_{amb} ; B) NPQ measured at 10s high light (NPQ_{high}); C) The high light-induced differences in NPQ ($NPQ_{high-amb}$); D) The difference in induced NPQ level at ambient PAR and the 10s recovery time in the dark ($NPQ_{amb-rec}$); E) The difference in induced NPQ level at 10s high PAR and the 10s recovery time in the dark ($NPQ_{high-rec}$); F) the NPQ level after 10s in the dark (NPQ_{rec}).

Figure 3B shows NPQ_t values measured at 10s full sunlight (NPQ_{high}). The NPQ light potential, or light-induced differences in NPQ ($NPQ_{high-amb}$) are shown in Fig. 3C. While $NPQ_{high-amb}$ was always positive, both $NPQ_{high-amb}$ and NPQ_{high} were suppressed at low PAR_{amb} or T_{leaf} . NPQ_t measured after the 10s dark recovery period (NPQ_{rec} , Fig. 3F) was consistently lower than NPQ_{amb} and NPQ_{high} . The difference between NPQ_{amb} and NPQ_{rec} ($NPQ_{amb-rec}$, Fig. 3D) ranged from slightly negative at low PAR_{amb} , where the majority of NPQ_{amb} was rapidly reversible, to about one at the higher PAR_{amb} , where about half of NPQ_{amb} was rapidly reversed.

Overall, these results indicate that the majority of NPQ_{amb} as well NPQ_{high} recovered within 10s of darkness and can likely be attributed to qE, and thus, under our conditions, qE is likely to be the most important form of NPQ. The residual, more slowly reversible, components reaching a little above 2 are likely to include qI or qZ [44,45], although the limited time frame for the protocol does not allow us to rule out contributions from longer-lived qE.

As with LEF, a simple linear effects model (Table S1B) showed strong interactions between T_{leaf} and PAR_{amb} , on NPQ_{amb} and the corresponding GMM analysis identified four clusters (Fig. S5). Cluster 1, which encompassed the lowest range of PAR_{amb} values, showed strong dependence on PAR_{amb} , with no significant dependence on T_{leaf} . The remaining clusters showed either dependence solely on T_{leaf} (Cluster 4) or codependence on PAR_{amb} and T_{leaf} (Clusters 2 and 3). Because GMM clustering suggests that T_{leaf} and PAR_{amb} can interact or act independently, depending on conditions, we excluded the linear effects models and focused on GMM for analyses of the remaining parameters.

For the analysis of NPQ_{high} (Fig. S6), we used the clusters found for NPQ_{amb} (Fig. S5), allowing us to directly compare changes in correlations among parameters within each

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3 cluster [40]. Cluster 1, which encompassed the lowest range of PAR_{amb} values, showed
4 strong dependence of NPQ_{high} on both PAR_{amb} and T_{leaf} . This pattern of dependencies was in
5 contrast to that for Cluster 1 for NPQ_{amb} , which showed dependence solely on PAR_{amb} . At a
6 higher range of PAR_{amb} (Cluster 3), NPQ_{high} showed significant dependence solely on T_{leaf} ,
7 again in contrast to the corresponding cluster for NPQ_{amb} , which showed dependencies on
8 both PAR_{amb} and T_{leaf} . Overall, compared to NPQ_{amb} , NPQ_{high} showed increased dependence
9 on T_{leaf} in all clusters, suggesting that it is more substantially controlled by metabolic or
10 physiological factors (see Discussion).
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19 **The redox state of Q_A**

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22 Figure 4A shows the dependencies of Q_A redox state (qL) on PAR and T_{leaf} . qL measured at
23 PAR_{amb} (qL_{amb} , Fig. 4A), was relatively constant (ranging from about 0.3 to 0.75) across
24 PAR_{amb} , with somewhat higher values at both extremes of PAR_{amb} . Lower leaf temperatures
25 appeared to be associated with lower qL values, over the entire range of PAR_{amb} , although
26 the effect was particularly pronounced at low light. By contrast, qL measured at 10s of high
27 light (qL_{high} , Fig. 4B) showed strong dependence on PAR_{amb} , ranging from near zero (fully
28 reduced Q_A) at low PAR_{amb} , to almost one (fully oxidised) at higher PAR_{amb} . Again, low T_{leaf}
29 appeared to correlate with lower qL_{high} throughout the range of PAR_{amb} . Strikingly, as shown
30 in Fig. 4C, the high light treatment induced two distinct effects: At low PAR_{amb} and/or T_{leaf} ,
31 it induced a net reduction of Q_A , while it had the opposite effect at higher PAR_{amb} and T_{leaf} .
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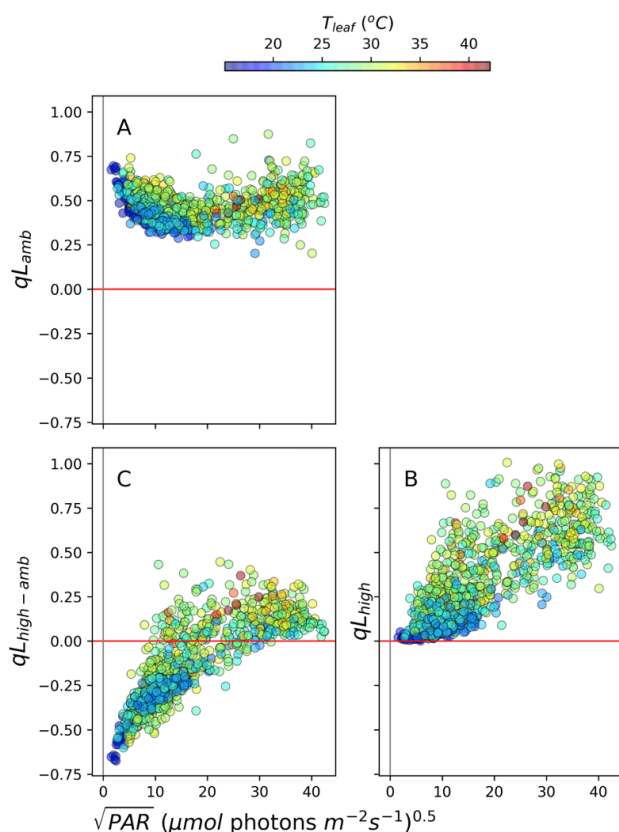


Figure 4. The light and temperature dependencies of the redox state of Q_A . The qL parameter, a measure of fraction of Q_A in its oxidized state, was measured as described in Material and Methods, under ambient light (qL_{amb} , Panel A), at 10s of high light (qL_{high} , Panel B), and the change in qL between high and ambient PAR ($qL_{high-amb}$, Panel C) as a functions of the square root of ambient PAR.

GMM clustering for qL_{amb} , PAR_{amb} and T_{leaf} (Fig. S7) identified four distinct clusters. In Cluster 2, which encompasses points at low PAR_{amb} , significant associations were observed only between qL_{amb} and PAR_{amb} . Clusters 1,3 and 4 (at higher PAR_{amb}) showed co-dependencies between qL_{amb} and both PAR_{amb} and T_{leaf} . GMM clustering for qL_{high} , PAR_{amb} and T_{leaf} showed five distinct clusters (Fig. S8). Clusters 1,2 and 5, which encompassed generally lower ranges for PAR_{amb} and T_{leaf} , showed qL_{high} dependencies on both PAR_{amb} and T_{leaf} . Clusters 3 and 4 (generally with higher PAR_{amb} and T_{leaf} values) showed only dependencies on T_{leaf} . The overall pattern of cluster behaviour was similar to that observed with respect to NPQ_{amb} and NPQ_{high} .

P700 redox state

Figure 5 shows the extent of oxidised P_{700}^+ (P^+), based on the DIRK of absorbance changes at 810 nm. P_{700}^+ at PAR_{amb} (P_{amb}^+ , Fig. 5A), after ten seconds of high light (P_{high}^+ , Fig. 5B) and the light-induced difference ($P_{high-amb}^+$, Fig. 5C). The extent of P_{amb}^+ was nearly linearly related to PAR_{amb} . Increasing the light resulted in higher P^+ values (P_{high}^+), indicating that, in all cases, PSI became more oxidised at high light. The extent of the light-induced oxidation was dependent on PAR_{amb} , with lower extents at low PAR_{amb} , and a peak at about 200-300 $\mu\text{mol photons m}^{-2} \text{s}^{-1}$. The decrease at higher PAR_{amb} was probably due to the accumulation of pre-oxidised P_{700} prior to the high light treatment.

The full extent of P_{high}^+ was relatively constant over the conditions, suggesting that high light was able to nearly fully oxidise P_{700} . However, there was a slight trend to lower P_{high}^+ at the highest PAR_{amb} or T_{leaf} , suggesting that total oxidizable PSI may have decreased at high light or temperatures, perhaps reflecting accumulation of PSI photodamage or electron sink limitations. Consistent with these general trends, GMM analyses of P_{amb}^+ , PAR_{amb} and T_{leaf} identified four distinct clusters (Fig. S9), with dependencies on either PAR_{amb} by itself (Clusters 3 and 4), or both PAR_{amb} and T_{leaf} (Clusters 1 and 2). GMM clustering for P_{high}^+ identified five distinct clusters (Fig. S10), that showed a positive dependency of P_{high}^+ on either PAR_{amb} (Cluster 1), or T_{leaf} (Cluster 5), or a small, negative dependence on T_{leaf} (Cluster 3).

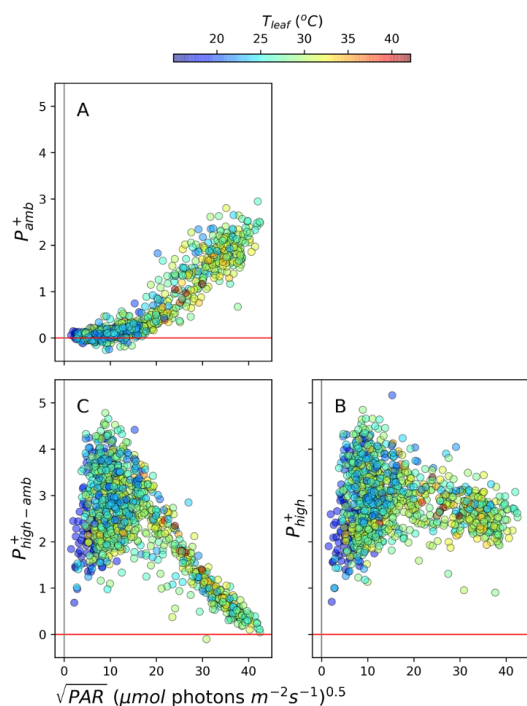


Figure 5. The light and temperature dependencies of the redox state of P_{700}^+ . The redox state of P_{700} was measured using DIRK at 810 nm absorbance change under ambient light (P_{amb}^+ , Panel A), at 10s of high light (P_{high}^+ , Panel B), and the change in P^+ between high and ambient PAR ($P_{high-amb}^+$, Panel C) as a functions of the square root of ambient PAR.

ECSt and thylakoid *pmf*

Figure 6 shows dependencies of relative thylakoid *pmf*, estimated by normalised ECSt measurements, at ambient PAR (ECSt_{amb}, Fig. 6A) and after 10s exposure to high light (ECSt_{high}, Fig. 6B). The high light-induced differences (ECSt_{high-amb}) are shown in Fig. 6C. ECSt_{amb} showed strong, positive correlations with PAR_{amb}, similar to the responses of NPQ_{amb} (Fig. 3A) and P_{amb}^+ (Fig. 5A). ECSt_{high} values were, in general, larger than ECSt_{amb}, resulting in positive values for ECSt_{high-amb}. At low PAR_{amb}, ECSt_{high} showed high variability, suggesting that the response is strongly dependent on other factors, but appeared to saturate (flatten) at higher PAR_{amb}. These behaviours were reflected in ECSt_{high-amb}, which showed strong variability at lower PAR_{amb} or T_{leaf}, peaked at about 50-100 $\mu\text{mol photons m}^{-2} \text{s}^{-1}$, and saturated at higher PAR_{amb}.

GMM analysis of ECSt_{amb} identified five distinct clusters (Fig. S11). The cluster at the lowest range of PAR_{amb} (Cluster 1) showed dependence primarily on PAR_{amb}. The remaining clusters showed positive correlations between ECSt_{amb} and PAR_{amb}, but negative correlations with T_{leaf}. By contrast, GMM of ECSt_{high} (Fig. S12) showed almost no dependence on either PAR_{amb} or T_{leaf}, except at the lowest PAR_{amb} (Cluster 1) which showed negative correlations with PAR_{amb} and positive correlations with T_{leaf}.

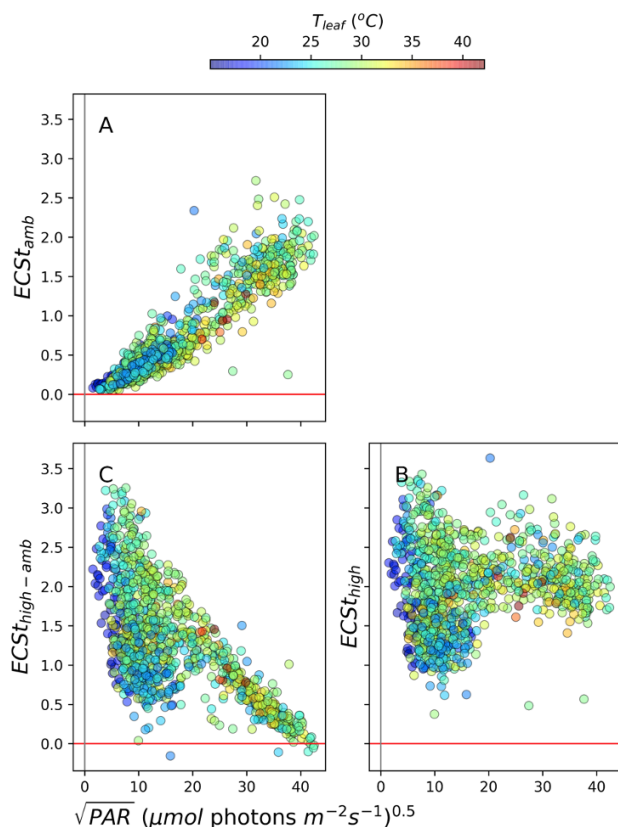


Figure 6. The light and temperature dependencies of the thylakoid *pmf* probed using ECSt signal. The *pmf* was measured using ECS under ambient light ($ECSt_{amb}$, Panel A), at 10s of high light ($ECSt_{high}$, Panel B), and the change in ECSt between high and ambient PAR ($ECSt_{high-amb}$, Panel C) as a functions of the square root of ambient PAR.

Discussion

Using PhotosynQ and MultispeQ to sample and resolve the effects of environmental fluctuations on photosynthetic processes.

The MultispeQ measurements described above were designed to explore the photosynthetic responses of plants in a natural, fluctuating environment. In this type of field experiment, it is not possible to control all variables. Rather, the strategy was to “sample” responses under as many conditions as practical, while recording key metadata so that subsequent analyses can assess the impacts of various environmental fluctuations. Thus the observed trends may

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3 reflect both primary and acclimatory factors that change (or accumulate) over different time
4 scales. Correlations that appear in such analyses can be used to test, at least to some extent,
5 certain models, though it is important to note that more controlled experiments will be
6 needed to fully determine cause-effect relationships, as discussed below.
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12 A major outcome of the experiment is that, despite the fact that measurements were made
13 over many plants, times etc., clear patterns of responses emerged that allow us to make
14 some broad conclusions about the responses of photosynthesis to ambient and rapidly
15 changing light. For example, the majority of NPQ_{high} was found, in general, to be rapidly
16 variable, suggesting that qE was the major contributor: At lower PAR_{amb} that majority of
17 NPQ_{high} was rapidly induced (see Fig. 3C), while at higher PAR_{amb} , pre-existing NPQ was
18 rapidly recoverable (Fig. 3E) at higher PAR_{amb} .
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26 Another important trend was the suppression of the light potentials of both LEF (Fig. 2) and
27 NPQ (Fig. 3) under some conditions, particularly under lower PAR_{amb} and/or T_{leaf} . Further,
28 strong decreases in LEF_{high} were not always accompanied by compensatory increases in
29 NPQ_{high} , implying that the productive and photoprotective light potentials can be
30 simultaneously suppressed under certain conditions, a situation that is likely to promote the
31 formation of reactive oxygen species and photodamage (see also below), with important
32 implications for understanding the environmental robustness of photosynthesis [46].
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40 **Disentangling interacting environmental impacts on photosynthetic processes.**

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43 A key challenge to the field experiment approach is in teasing apart effects from different
44 environmental factors, especially considering that such factors may be codependent or
45 interact with each other in complex ways. For example, in visual inspection, most of
46 parameters show apparent dependencies on both PAR_{amb} and T_{leaf} (e.g. Figs. 2-6) but,
47 because increases in T_{leaf} are often correlated with increases in PAR_{amb} , the effects of the
48 two parameters may have been coincidental. It may also be that the environmental
49 parameters interacted in complex ways, e.g. high PAR_{amb} may have exacerbated the effects
50 of low T_{leaf} . To address these issues, we applied an approach based on GMM to identify
51 clusters representing distinct interactions among parameters. The approach is unsupervised,
52 thus eliminating potential bias, while allowing us to test for changes in the environmental
53 dependencies among multiple environmental parameters (Figs. S3-S12).
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5 Analysis of GMM clusters implied that most parameters were dependent on both PAR_{amb}
6 and T_{leaf} , and at least under some conditions these effects are independent, or that one of the
7 two factors predominates. Thus, the effects cannot be explained simply by coincidences
8 between increased PAR and temperatures. Moreover, the non-rectilinear shapes of the
9 clusters suggest that the effects of PAR_{amb} and T_{leaf} were interactive, e.g., changes in T_{leaf}
10 modulated the effects of PAR_{amb} and vice versa. Overall, these interactions are in line with
11 well-known temperature and PAR dependence of photosynthesis, but this type of analyses
12 can reveal the specific combination of conditions that induce distinct behaviours, allowing
13 for assessments of the involvement of specific mechanisms (see below) and to identify
14 genotypic or management impacts on crop resilience and productivity.
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24 At low PAR_{amb} , we expect steady-state photosynthesis to be predominantly light-limited,
25 and thus the effects of T_{leaf} should be low. As light increases, downstream biochemistry
26 should become increasingly limiting. Because downstream energy storage and metabolic
27 processes are likely to be more temperature dependent than photochemistry, this shift may
28 allow us to distinguish between these types of limitations. Such behaviours are apparent in
29 many of the measured parameters, e.g., LEF_{amb} , which was not substantially dependent on
30 T_{leaf} at low PAR_{amb} , but became codependent on PAR_{amb} and T_{leaf} at higher PAR_{amb} (Figs. 2A,
31 S3), consistent with a progressive shift from light-limitation to assimilation-limitation.
32 Similarly, NPQ_{amb} was solely dependent on PAR_{amb} in the cluster at low PAR_{amb} , but became
33 increasingly dependent on T_{leaf} as PAR_{amb} increased (Fig. 3A). This shift is consistent with a
34 control of NPQ_{amb} by PAR (at low PAR_{amb}) and downstream metabolic processes,
35 particularly at higher PAR_{amb} , e.g., due to regulation of the ATP synthase activity or cyclic
36 electron flow (CEF) [47].
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48 By contrast, LEF_{high} and NPQ_{high} showed much greater dependence on T_{leaf} and these
49 differences were more pronounced when the high light was imposed on leaves at low
50 PAR_{amb} and T_{leaf} , i.e. the opposite of what was seen for LEF_{amb} and NPQ_{amb} . Interestingly,
51 the LEF_{high} rates achieved in leaves exposed to lower PAR_{amb} were strongly suppressed
52 below the maximum LEF_{amb} values measured at higher PAR_{amb} (compare Figs. 2A and B),
53 This behavior suggests that the suppression of LEF_{high} occurs when abrupt increases in light
54 overwhelm the activation of downstream energy storage and metabolic processes. This is
55 generally consistent with observations that the activities of metabolic enzymes are regulated
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3 to match the availability of energy from the light reactions, which involve large suite of
4 co-regulatory processes, as extensively reviewed elsewhere, e.g. [16,47–53], but that these
5 responses lag behind the changes in light. The *in situ* light potential measurements afforded
6 by MultispeQ show that these situations are very likely to occur under many field
7 situations.
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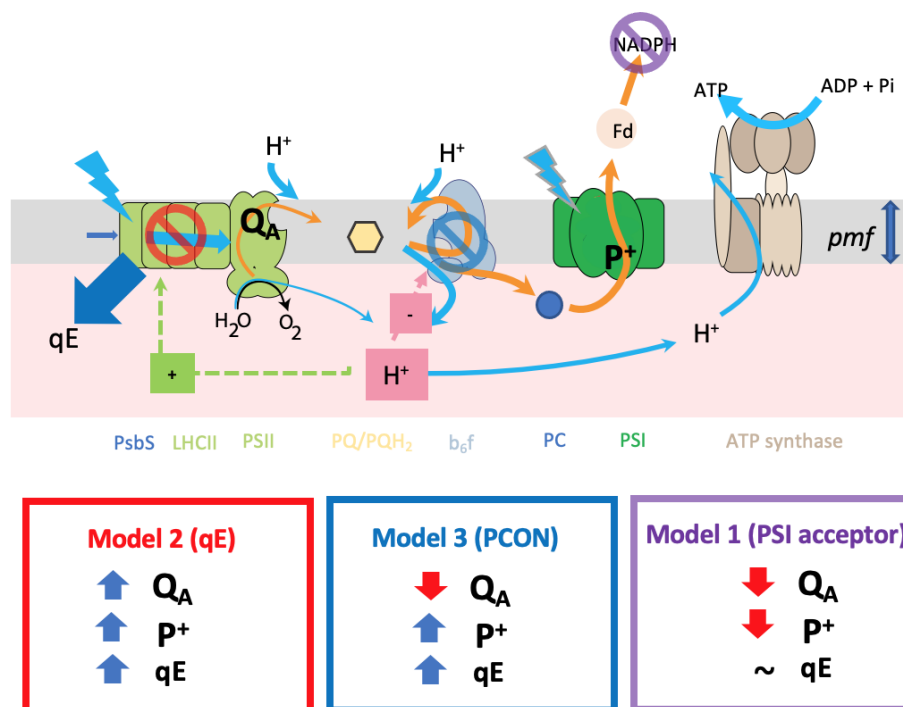
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13 These results also imply that accurate estimates of LEF, NPQ and other photosynthetic
14 parameters will require measurements under ambient light, because sudden changes in PAR
15 can lead to severe perturbations in photosynthetic limitations or regulation. Attempts to
16 “simplify” field experiments by setting PAR to some constant value will lead to strong
17 artifacts. Such effects are vividly demonstrated by the opposite dependencies of $\Phi_{2_{amb}}$ and
18 $\Phi_{2_{high}}$ on PAR_{amb} (Fig. 2D), and validate the use of the PAR matching feature of the
19 MultispeQ instrument. It is important to keep in mind that the rates of acclimation may vary
20 substantially between species, and that these may be assessed by performing more intensive
21 experiments with variable high light and dark recovery times.
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30 **Mechanisms for controlling the light potentials of LEF and NPQ using MultispeQ** 31 **field data.** 32 33

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36 The rapid reversal of NPQ_{amb} and NPQ_{high} over 10 s of dark indicated that, under our
37 conditions, NPQ is predominantly in the form of qE (Fig. 3B and C), and thus dependent on
38 lumen acidification and subsequent pH-dependent responses. Lumen acidification can be
39 controlled by changes in proton influx (through changes in LEF and CEF), proton efflux
40 through the ATP synthase and the partitioning of *pmf* into electric field ($\Delta\psi$) and ΔpH
41 components, which in turn, are impacted by metabolic status, as proposed earlier [15,38].
42 Here, we explore the possible mechanistic bases for these effects, by comparing the
43 correlations among MultispeQ measurements.
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51 Scheme 1 illustrates three basic mechanistic models describing proposed processes that can
52 limit the light potentials of photosynthetic and photoprotective mechanisms. The models
53 make qualitative predictions about how the actions of each mechanistic model will impact
54 correlations between measured photosynthetic parameters, and thus can be used as a
55 framework for interpreting the field data introduced in Results. The expected effects on the
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measured parameters are summarised in Scheme 1, which shows specific effects of each model.



Scheme 1: Models for limitations to LEF and NPQ light potentials.

Model 1: PSI acceptor-side limitations (Scheme 1, Model 1) where lack of NADP⁺, ferredoxin or other PSI acceptors prevent further LEF. We expect this limitation to result in accumulation of electrons throughout the electron transfer chain, thus resulting in net reduction of Q_A (decreasing qL) and P₇₀₀⁺ (decreasing the 810 nm absorbance signal). The decreases in proton fluxes associated with backup of electrons may, in addition, prohibit rapid, light-induced increases in *pmf*, lumen acidification and qE responses.

Model 2: Increased NPQ (Scheme 1, Model 2). Increased NPQ should decrease delivery of excitation energy to PSII (but not to PSI), resulting in net oxidation of Q_A (increasing qL) and P₇₀₀⁺ (increased 810 nm DIRK signal). Under some conditions, the NPQ will be rapidly induced by increased *pmf* and lumen acidification followed by activation of qE, which should be visible as increased NPQ_{high-amb}. Under other conditions, e.g. at higher PAR_{amb}, NPQ may already have been induced. If this NPQ is in the form of rapidly-reversible qE, it

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3 should substantially decay during the 10-s dark recovery period, resulting in increased
4 NPQ_{high-rec}. More slowly-induced or relaxing forms of NPQ, including qI, qZ and long-lived
5 qE, may be also present prior to and throughout the experiment. The forms should register
6 as increases in NPQ_{rec}, but not in NPQ_{high-amb} or NPQ_{high-rec}, but given that the high light and
7 recovery periods were only 10s long, our results do not allow us to distinguish among these
8 possible forms.
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15 **Model 3: Photosynthetic control (PCON, Scheme 1, Model 3).** PCON results from the
16 slowing of PQH₂ oxidation at the cytochrome *b_f* complex as the lumen becomes acidified.
17 If PCON occurs without activation of qE, we expect a net reduction of Q_A (decreasing qL)
18 but a net oxidation of P⁺ (increasing the 810 nm absorbance signal).
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24 The qE and PCON models can be further subdivided [18,54]. In most cases, we expect
25 lumen acidification accompanied by elevated *pmf*, reflected in an increased ECSt signal,
26 which can be induced by increased proton influx into the lumen, due to increased LEF,
27 increased CEF, or decreased conductivity of the thylakoid to protons (g_{H^+}) by slowing the
28 ATP synthase. Alternatively, lumen acidification can also be associated with an increase in
29 the fraction of *pmf* that is stored as ΔpH , by controlling the flow of counterions across the
30 thylakoid membrane, altering the partitioning of *pmf* in ΔpH and $\Delta\psi$. In this case,
31 acidification may occur with little or no increases in total *pmf*, or the rates of proton influx,
32 though the current field-based data do not allow us to directly distinguish these possibilities.
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43 These models, while not mutually exclusive, will tend to counteract each other, at least
44 within a particular leaf. For instance, PSI acceptor side limitations will tend to inhibit
45 electron flow, thus decreasing proton flux and *pmf* generations. On the other hand, the
46 generation of *pmf* will tend to slow electron flow (through Models 2 or 3), thus preventing
47 the buildup of electrons on PSI electron acceptors. However, it is important to note that, in a
48 survey type experiment like ours, photosynthesis in different leaves may be limited by
49 distinct processes, and thus any collection of samples may reflect various combinations of
50 the above models.
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58 **Testing models for limitations in light potentials.**
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By plotting MultispeQ parameters against each other, we can test for more detailed patterns of behaviours predicted by the above models. Figure 7 shows that $P_{\text{high-amb}}^+$ (high-light-induced P_{700} oxidation) was positively correlated with light-induced increases in pmf ($ECSt_{\text{high-amb}}$). Under all conditions, increasing PAR from PAR_{amb} to PAR_{high} resulted in a net oxidation of P_{700} , i.e., $P_{\text{high-amb}}^+$ was consistently positive. This behaviour is consistent with Models 2 (NPQ) or 3 (PCON), both of which predict a decrease in delivery of electrons from PSII to PSI. By contrast, we did not see evidence for high light-induced net reduction of P_{700}^+ , i.e., values of negative $P_{\text{high-amb}}^+$, implying that Model 1 was not a major limitation to light potential. This does not exclude Model 1 from limiting photosynthesis in different species and conditions, as has been proposed to be important in chilling sensitive plants [55]. However, the apparent avoidance of Model 1 (or prevalence of Models 2 and 3) behaviour may reflect the “tuning” of the light reactions to prevent the accumulation of reduced electron acceptors of PSI associated with photodamage [23], and the associated O_2 caused by buildup of electrons on PSI [56].

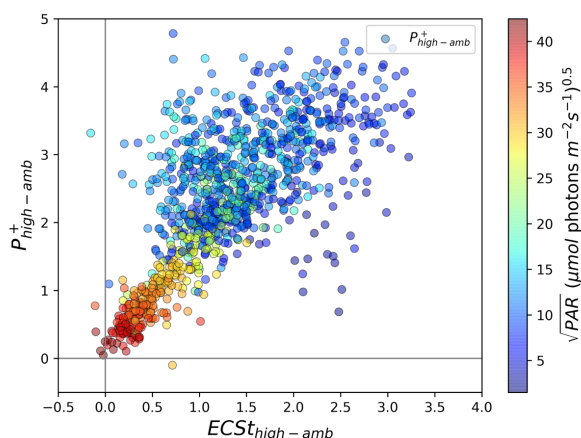


Figure 7. The relationship between light-induced thylakoid pmf and changes in P_{700} redox state. Changes in the thylakoid pmf ($ECSt_{\text{high-amb}}$) were estimated using the $ECSt$ parameter, and changes in P_{700}^+ were measured using the absorbance changes at 810 nm, as described in Materials and Methods, under ambient light ($ECSt_{\text{amb}}$, P_{amb}^+) and after 10s of high light ($ECSt_{\text{high}}$, P_{amb}^+). The coloration of the points was set to a function of the square root of ambient PAR (PAR_{amb}).

Overall, the behaviours seen in Fig. 7 are consistent with restrictions in electron flow to PSI imposed by increases in pmf , most likely through the acidification of the thylakoid lumen. In the case of Model 2 (rapid NPQ), this would be related to the induction of qE, while in Model 3 (PCON), this could be related to slowing of electron flow at the cytochrome b_6/f complex.

Figure 8A further investigates this behaviour by plotting the dependence of high light-induced changes in P_{700}^+ ($P_{\text{high-amb}}^+$) with changes in Q_A redox state ($qL_{\text{high-amb}}$). The expected theoretical behaviours for the three models are indicated by the coloured boxes in the figure, and can be related to Models 1-3 in Scheme 1:

- **Model 1** (violet box) predicts net **reduction** of P_{700} ($P_{\text{high-amb}}^+ < 0$) and net **reduction** Q_A ($qL_{\text{high-amb}} < 0$)
- **Model 2** (red box) predicts net **oxidation** of P_{700} ($P_{\text{high-amb}}^+ > 0$) and net **oxidation** Q_A ($qL_{\text{high-amb}} > 0$)
- **Model 3** (blue box) predicts net **oxidation** of P_{700} ($P_{\text{high-amb}}^+ > 0$) but net **reduction** Q_A ($qL_{\text{high-amb}} < 0$)

We observe behaviours consistent with both Models 2 and 3, suggesting that the behaviour of the system changed with conditions. Note that the boxes in Figure 8A represent “pure” behaviours, and it is possible that the effects of a particular mechanism may be intermediate, e.g., the responses may be limited by a combination of reduction of Q_A and increased NPQ.

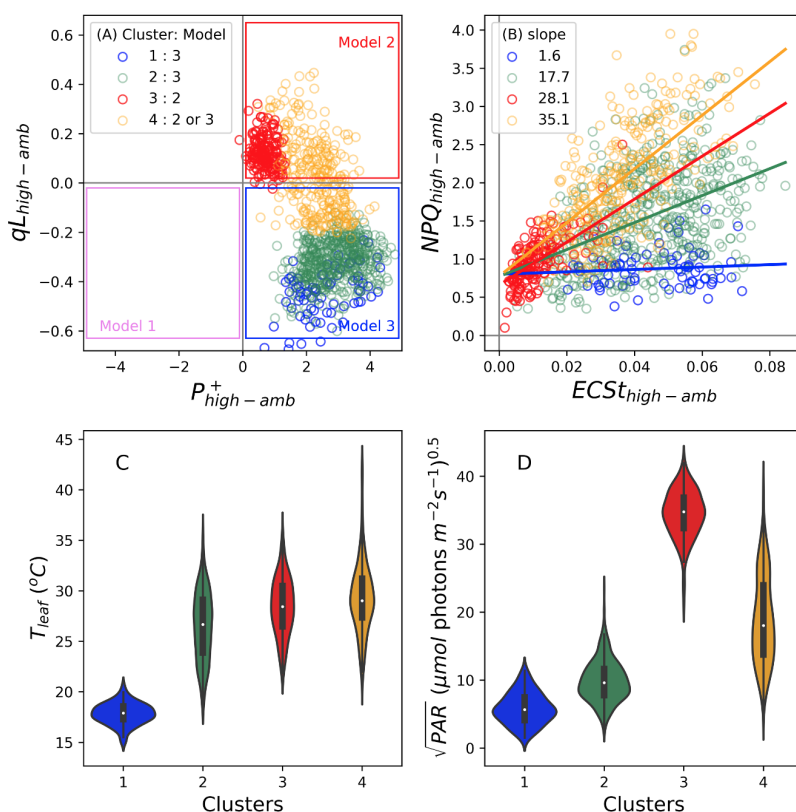


Figure 8. The relationships between light-induced changes in Q_A redox state and P_{700} redox state (Panel A), and between rapidly inducible NPQ and thylakoid pmf (Panel B) and the leaf temperature (Panel C) and PAR (Panel D) dependencies of Gaussian Mixture Models (GMM) clusters. Changes in P_{700}^+ ($P_{high-amb}^+$), Q_A redox state ($qL_{high-amb}$), rapid changes in NPQ ($NPQ_{high-amb}$) and thylakoid pmf ($ECSt_{high-amb}$) were measured as described in Materials and Methods. Data were clustered using the GMM approach described in the text, resulting in four distinct clusters, designated by the blue, green, red and orange symbol colors (see legend in Panel A). In Panel B, the slopes for the relationship between $NPQ_{high-amb}$ and $ECSt_{high-amb}$ were estimated by linear regression (slopes for clusters 1,2,3 and 4 were estimated to be 1.6, 17.7, 28.1 and 35.1, respectively). Panels C and D show distributions of leaf temperatures (T_{leaf} , Panel C) and square root of ambient PAR for each cluster in Panels A and B.

Figure 8B plots the dependence of $NPQ_{high-amb}$, which can be attributed to light-induced qE changes, on light-induced pmf changes ($ECSt_{high-amb}$). A generally positive correlation was observed between $NPQ_{high-amb}$ and $ECSt_{high-amb}$, but with high variability, especially at higher values. Applying the clustering obtained for Fig. 8A on top of the data in Fig. 8B, we see that this variability can be explained by the environmental conditions and the modes of behaviours. Specifically, we see clear evidence for condition-dependent suppression of rapid activation of qE in response to increases in pmf . Particularly, the sensitivities of $NPQ_{high-amb}$ to $ECSt_{high-amb}$, as indicated by the slopes in Fig. 8B, were smallest in Clusters 1 (slope ~ 1.6) and 2 (slope ~ 17.7), which comprise those with Model 3-like behaviour and occurred at low T_{leaf} and PAR_{amb} values. Higher sensitivities of $NPQ_{high-amb}$ to $ECSt_{high-amb}$ were seen for Clusters 3 (slope ~ 28.1) and 4 (slope ~ 35.1), which comprised those associated with Models 2 and intermediate, and occurred at higher T_{leaf} and PAR_{amb} values.

To assess what controlled the switch between Models 2 and 3, we performed GMM (using $qL_{high-amb}$, $P_{high-amb}^+$, T_{leaf} as inputs). Four distinct clusters were observed (see symbol colours, Fig. 8A). Intercluster comparisons show that points in Clusters 1 and 2 fell exclusively in the region predicted for Model 3. Cluster 3 fell entirely within the region predicted for Model 2. Cluster 4 extended between these regions, possibly indicating contributions from both mechanisms. The clusters falling in the Model 3 region were associated with relatively low T_{leaf} (Fig. 8C) and PAR_{amb} (Fig. 8D), compared to those associated with Model 2 or intermediate behaviours, suggesting that Model 2 prevailed at higher T_{leaf} and/or PAR_{amb} , while Model 3 prevailed at lower values. Within the GMM clusters (Fig. S13), $qL_{high-amb}$ was dependent predominantly on T_{leaf} (Cluster 3), PAR_{amb} (Cluster 1), or both (Clusters 2 and 4). This dependence suggests that T_{leaf} and PAR_{amb} acted either independently or

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3 cooperatively, depending on conditions, affecting the propensity for photosynthesis to adopt
4 Model 2 or 3 behaviours.
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8 The data in Figs. 8 show that, at lower T_{leaf} and PAR_{amb} , qE activation was suppressed
9 despite light-induced increases in pmf , and that this behaviour was associated with
10 accumulation of electrons on Q_A but oxidation of P_{700} (Fig. 8A), suggesting that, under
11 these conditions, light-induced increases in ΔpH caused slowing of the cytochrome bf
12 complex (PCON), but that the qE response lagged behind or was completely suppressed,
13 leading to Model 3 behaviour.
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20 It has been shown that the lumen pH-dependencies of qE and PQH_2 oxidation by the
21 cytochrome bf complex are tightly coordinated, so that increased lumen acidity activates
22 photoprotection prior to PCON, presumably to prevent the accumulation of reduced Q_A
23 [54]. However, these experiments were performed under more slowly-changing (near
24 steady-state) conditions in the laboratory, and our results suggest that this coordination can
25 be defeated under real world conditions in the field, especially when T_{leaf} is low and PAR
26 fluctuates rapidly. This discoordination can have strong implications for photodamage, as it
27 has been shown that high thylakoid pmf can greatly accelerate PSII recombination
28 reactions, especially when Q_A is reduced, leading to $^1\text{O}_2$ production [28,29]. It thus seems
29 reasonable to suggest that the shift from qE to PCON at low T_{leaf} will increase the rates of
30 photodamage.
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41 There are several possible mechanisms by which the response of qE can be uncoupled from
42 increased pmf . Longer-term dependencies of NPQ on temperature have been reported under
43 both field [57–59] and laboratory [60][61] conditions. The current work shows effects on
44 rapid NPQ and LEF changes, which can be related to distinct mechanistic models. For
45 example, it is known that the xanthophyll cycle is strongly temperature dependent, though
46 the general observation is that zeaxanthin tends to accumulate at lower temperatures due to
47 a slowing of the epoxidation of zeaxanthin [60][61,62]. Interestingly, we would expect the
48 accumulation of zeaxanthin to augment, rather than suppress qE responses as we have
49 observed in the current results. Lumen acidification may also be rate limiting for formation
50 of qE. While rapid increase in light can result in nearly instantaneous increases in $\Delta\psi$,
51 formation of ΔpH and lumen acidification requires counterion transport processes, which
52 tend to be slow, and thus lumen acidification lags behind [25,29], and it is possible that this
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3 process is substantially slowed at low temperature. Other possible limitations include
4 temperature-dependence of conformational rearrangement of antenna complexes following
5 protonation of PsbS [63,64], which in turn may be related to the interactions among
6 thylakoid proteins, lipids and ultrastructure [12,44,65,66]. The current data does not allow
7 us to discriminate between these models with the current data set, but the work suggests
8 conditions and species under which such limitations occur, and how they may impact plant
9 productivity or resilience.
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17 **Conclusion: Current limitations and prospects for open science-led efforts to**
18 **understand and improve photosynthesis.**
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22 There are intense, ongoing efforts to improve photosynthesis, yet the importance of the
23 responses of photosynthesis under fluctuating, real world conditions are just now being
24 recognised. In particular, we lack understanding of the extents and impacts of these
25 responses, as well as their mechanisms and genomic control, which will be critical to
26 achieving field-relevant improvements in efficiency and robustness, especially in a
27 changing environment.
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34 Here, we demonstrate methods and tools to assess the light responses of photosynthetic
35 processes under real world conditions, and use them to explore the factors that limit the
36 capacity of plants to utilise or dissipate rapidly increased PAR. A major outcome is that,
37 despite the complexities of field environments, clear behavioural patterns can be resolved,
38 as long as the experiment contains a sufficient number of points taken over a large
39 environmental space, and that includes both environmental metadata. Such combinations of
40 information allow for the generation and testing of specific hypotheses. For example, we
41 observed no evidence for Model 1 behaviour in the current study, but we do not exclude the
42 possibility in different species and/or different environmental conditions. The rapid
43 measurements allowed us to test for various models over broad-scales by looking for
44 internally-consistent relationships among the various measured parameters. Further, while
45 we surmised (above) that Model 3 type behaviour would likely lead to photodamage, we do
46 not have independent endpoint measurements (e.g., yield, growth rates etc.) to validate that
47 the propensity for Model 3 behaviour has long-term consequences. The models are not
48 exclusive, and there will almost certainly be cases, e.g., Cluster 4 in Fig. 8, where
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3 intermediate behaviours will be apparent, either because of co-limitations among multiple
4 processes or heterogeneity between chloroplasts in the leaf samples.
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8 We also emphasise that the data presented here was intended to introduce the approaches
9 and methods, and thus leaves a number of questions unanswered, but sets up the approach
10 to further study. The origins of these effects may include several classes of processes
11 [31,67], that may differ under different conditions [68], including induction of downstream
12 assimilatory reactions and metabolic pools [69,70], downstream sink reactions [71], redox
13 regulation [72,73], balancing between the production and consumption of ATP and NADPH
14 [1,49], ion homeostasis and regulation of thylakoid *pmf* [25,74], low stomatal aperture that
15 may lead to transient depletion of internal CO₂ levels . Distinguishing these will likely
16 require more detailed phenotyping and biochemical [10,56,75], modeling [31] and
17 genomics and genetics approaches [76].
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27 The accessibility of the tools should allow larger numbers of researchers to answer these
28 types of questions over a broader set of results, as will be presented in an upcoming paper.
29 This approach was made possible by the combination of several open science advances.
30 Collation of large amounts of data and metadata through the MultispeQ and PhotosynQ
31 platforms [33], allowing us to explore the interdependencies of multiple phenotypes and
32 environmental conditions (metadata). The GMM methods allowed us to explore the
33 interactions among multiple environmental parameters and photosynthetic phenotypes, and
34 test for the participation of distinct mechanistic models to explain the limitations to
35 photosynthesis under field conditions, leading to the identification of distinct limitations in
36 the rapid activation of NPQ and LEF at low temperature. Finally, making all tools,
37 protocols and analytical methods available in directly usable forms, the project can be
38 readily expanded to include multiple environments and species, as well as alternative
39 models.
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Data accessibility. Primary data is available on the photosynq.org site under the project “rapid-ps-responses-pam-ecst-npqt-mint-dmk”. Data cleaning and analysis code is available in a GitHub repository (<https://github.com/protonzilla/Light-Potentials-in-Field>).

Competing interests. D.M.K. and S.K. are co-founders of PhotosynQ which maintains the PhotosynQ platforms and distributes and maintains the MultispeQ instruments. The current project was performed independently with no funding to or from the PhotosynQ organization.

Author contributions. A.K. and D.M.K. designed the experiments. A.K. and H.T. conducted experiments. A.K., A.C., S.K. and D.M.K. analyzed data. A.K., A.C., S.K., T.M. and D.M.K. contributed to the interpretations of data and writing manuscript.

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