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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Summary. The responses of plant photosynthesis to rapid fluctuations in environmental conditions are thought to be critical for efficient capture of light energy. Such responses are not well represented under laboratory conditions, but have also been difficult to probe in complex field environments. We demonstrate an open science approach to this problem that combines multifaceted measurements of photosynthesis and environmental conditions, and an unsupervised statistical clustering approach. In a selected set of data on mint (*Mentha* sp.), we show that the “light potential” for increasing linear electron flow (LEF) and nonphotochemical quenching (NPQ) upon rapid light increases are strongly suppressed in leaves previously exposed to low ambient PAR or low leaf temperatures, factors that can act both independently and cooperatively. Further analyses allowed us to test specific mechanisms. With decreasing leaf temperature or PAR, limitations to photosynthesis during high light fluctuations shifted from rapidly-induced NPQ to photosynthetic control (PCON) of electron flow at the cytochrome *b₆f* complex. At low temperatures, high light induced lumen acidification, but did not induce NPQ, leading to accumulation of reduced electron transfer intermediates, a situation likely to induce photodamage, and represents a potential target for improving the efficiency and robustness of photosynthesis. Finally, we discuss the

implications of the approach for open science efforts to understand and improve crop productivity.

Keywords: nonphotochemical quenching (NPQ), photodamage, proton motive force, qE, photosynthetic control, unsupervised learning

Introduction

While oxygenic photosynthesis supplies energy to drive essentially all biology in our ecosystem, it involves highly energetic intermediates that can generate highly toxic reactive oxygen species (ROS) that can damage the organisms it powers [1]. Thus, the energy input into photosynthesis must be tightly regulated by photoprotective mechanisms that act at several key steps in the light reactions. The balance and kinetics of this regulation is an active target for crop improvement.

One class of photoprotective processes, known as nonphotochemical quenching (NPQ), dissipates absorbed light energy as heat, thus diverting energy away from photosystem II (PSII) [2], decreasing the accumulation of reactive intermediates. This photoprotective capacity comes at the cost of decreased photochemical efficiency, and thus the organisms must regulate NPQ to balance the avoidance of photodamage with efficient energy conversion [3,4]. There are several forms of NPQ that differ in their mechanisms and rates of activation and deactivation. The most rapid NPQ form is qE, which is activated by acidification of the thylakoid lumen by the proton gradient (ΔpH) component of the thylakoid proton motive force (pmf) [2]. Lumen acidification activates the violaxanthin deepoxidase or VDE [5–8] resulting in the conversion of violaxanthin (Vx) to antheraxanthin (Ax) and zeaxanthin (Zx); and protonation of PsbS, an antenna-associated protein required for qE [2], which appear to act cooperatively in setting the extent of qE. The conversion of Vx to Ax and to Zx is typically much slower than the rapidly reversible protonation of PsbS [2], and during prolonged illumination, the responses of qE will likely be limited by the rate of acidification and de-acidification of the thylakoid lumen, which are, in turn, governed by ion movements in the chloroplasts [9–11]. Slower forms of NPQ have also been demonstrated [12], including qI, which is related to the photodamage and repair of photosystem II (PSII) or qZ, which related to the accumulation of Zx

(independently from qE) [13], qH, related to cold and high light stress [13], and qT, related to antenna state transitions [14].

The acidification of the thylakoid lumen also controls electron transfer at the cytochrome *b₆f* complex, a process called photosynthetic control (PCON) [15–20], which prevents the buildup of electrons on the acceptor side of photosystem I (PSI) that can lead to photodamage [15,21–23]. Interestingly, PCON and qE (both responses to lumen acidification) are expected to have opposing effects on Q_A redox state. High levels of PCON in the absence of qE would lead to accumulation of plastoquinol (PQH₂) and the reduced form of the PSII electron acceptor, Q_A⁻, which can potentiate photodamage. Thus, these two processes must be tightly coordinated, with qE being activated at lumen pH somewhat less acidic than PCON [15].

Plants in natural environments are exposed to rapidly changing environmental conditions, especially light which can change by orders of magnitude in less than a second. It has become clear that rapid and unpredictable fluctuations in light intensity can be more damaging than more gradual changes [22,24–31]. This sensitivity can partly be related to the buildup of reactive redox intermediates and thylakoid *pmf*, which can occur following low-to-high light transitions much more rapidly than the activation of photoprotective NPQ and PCON, leaving the photosynthetic apparatus prone to photodamage. Also, the slow recovery of NPQ following a decrease in light intensity can lead to substantial losses of photosynthetic efficiency [32]. Recently, it has been reported that engineering plants with increased expression levels of VDE and zeaxanthin epoxidase (ZE), resulted in accelerated formation and reversal of qE accompanied by increased plant productivity [3], suggesting that it may be possible to increase yield in crops by modifying photosynthetic regulatory responses.

On the other hand, we lack comprehensive surveys of the range of natural response of photosynthesis to real environmental fluctuations, in part because of a lack of deployable scientific equipment and methods to probe these processes in the field. Consequently, it has not been possible to assess the mechanistic bases of extant natural variations in these processes, their possible benefits or tradeoffs, or which of these may be most useful for crop improvement.

Here, we introduce a method and proof-of-concept field data results to address the following questions: Can we assess the extent of natural variations in rapid responses to fluctuations in photosynthetically-active radiation (PAR) intensity for both electron flow and photoprotection? How do these limitations depend on environmental conditions? What are the mechanisms that underlie these variations in responses to rapidly fluctuating light in the field?

Here, we introduce an approach to both measure and analyse these variations, focusing on one species, *Mentha* sp., under a limited set of conditions, and applied these to testing among a set of mechanisms that can be distinguished based on a range of optical measurements available using the MultispeQ 2.0 device, including: 1) PSI acceptor-side limitations to electron transfer; 2) Increased NPQ which limits the input of light energy into photosystem II (PSII); and 3) Photosynthetic control in which acidification of the lumen slows electron transfer at the level of plastoquinol (PQH₂) oxidation by the cytochrome *b₆f* complex.

The results show that the approach can effectively be used to assess the range of variations in ‘light potentials,’ the extent to which increased light leads to increased photosynthetic responses, under field conditions, as well as to test specific hypothetical models, setting up a broad-scale, multiple participant, open science approach to exploring the responses across multiple species, genotypes and environments. The results also reveal, at least in *Mentha*, unexpected leaf temperature-dependent limitations in the rapid formation of NPQ that result in the accumulation of reduced PSII electron acceptor, Q_A and a high thylakoid *pmf*, conditions likely to promote the formation of reactive oxygen species.

Materials and Methods

Plants and leaf sampling.

Measurements were made in a population of *Mentha* sp. (mint) plants that have been maintained at the MSU Horticulture Gardens for at least 10 years. The GPS locations of all measurements are included in the online data set (<https://photosynq.org/projects/rapid-ps-responses-pam-ecst-npqt-mint-dmk>). Although it was not practical to exhaustively capture the lifecycle of the plants, the experimental

strategy sampled a sufficiently wide range of conditions to allow clear patterns emerged in the relationships between phenotypes and environmental parameters, as described below. The experiment took place over a nine-day experimental window (Figure S1A), sampling a range of times of day, temperatures etc. (Fig. S1B). Measurements were made at multiple, alternating canopy levels and positions (subjectively at high, middle and low canopy levels) from early morning, though later afternoon (Fig. S1B), and at multiple locations across the plots on each day.

Measurements of photosynthetic and related parameters using MultispeQ 2.0.

Optical measurements were made using MultispeQ V2.0 hand-held instruments (<https://photosynq.com>), based on that presented by Kuhlert et al [33] and calibrated using the CaliQ calibration system (<https://photosynq.com/caliq>). The Light Potential (LP) protocol used in the experiments can be found in the online project information ([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 protocol was designed to strike a balance between needs for sampling large numbers of leaves, the desire for detailed spectroscopic measurements and the length of time the plant could be exposed to increased or decreased PAR. The full protocol, with measurements at ambient, after 10 s full sunlight and 10 s dark required about 35-40 s, at the limit of the time scale over which most researchers could steadily clamp a leaf in the instrument. The implications of the 10 s illumination and recovery time are discussed in the Results and Discussion sections.

In the first stage of the protocol (Fig. 1A), the MultispeQ was programmed to continuously (at about 5 Hz) measure photosynthetically-active radiation (PAR) and reproduced these levels using a red actinic LED (655 nm emission peak) illuminating the adaxial surface of the leaf. When the MultispeQ detected that a leaf was clamped in the chamber, a series of measurement sequences were initiated. After a few seconds of illumination at ambient PAR (PAR_{amb}) to allow for settling and setting of gains, the first set of measurements was made, estimating at PAR_{amb} LEF (LEF_{amb}), NPQt (NPQ_{amb}) and other photosynthetic parameters (Fig. 1B).

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_i after relaxation (NPQ_{rec} , Fig. 1F). Environmental parameters including PAR, temperature, humidity, leaf temperature, leaf

angle and GPS location were measured either prior to or following the physiological measurements.

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

Environmental conditions during light potential measurements in the field.

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

Data calculations and cleaning

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

Results

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

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

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 ($LEF_{\text{high-amb}}$); D) The PSII quantum efficiencies measured under ambient PAR ($\Phi_{II2\text{amb}}$, points colored by T_{leaf}) and at 10s high light ($\Phi_{II2\text{high}}$, 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 $LEF_{\text{high-amb}}$ 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 ($LEF_{\text{high-amb}}$) 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_{II2} , Fig. 2D). Φ_{II2} at PAR_{amb} ($\Phi_{II2\text{amb}}$) were highest at low PAR_{amb} , and progressively saturated as light was increased. The opposite behavior was seen with Φ_{II2} 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].

Clusters obtained through GMM have both within cluster (intracluster) and between cluster (intercluster) variations. In order to test for intercluster variation, we used the clustering assignment obtained for one phenotype and applied it on other phenotype(s). Here we want to investigate what would be the distinctive behaviour of different phenotypes if we have used the same configuration. Using the same set of cluster assignments to different phenotypes, one might be skeptical of the clustering behaviour as phenotypes interact differently with PAR_{amb} and T_{leaf} . In that case, we might not be able to directly compare the inter cluster behaviours of phenotypes. To mitigate this issue, we use the GMM clustering as a tool to create a “baseline” clustering configuration for one phenotype and use that configuration over another phenotype. We set up our hypothesis as two phenotypes are similar under the same configuration against they are not. If the interaction pattern of one phenotype with PAR_{amb} and T_{leaf} changes over the other phenotype, we reject our hypothesis and imply that different configurations of PAR_{amb} and T_{leaf} interact differently with phenotypes. By doing this we are able to disentangle the effect of PAR_{amb} and T_{leaf} and infer regarding the intracluster variations as to be a key element to determine variations in the interactions between parameters and variations in environmental conditions, e.g., to assess if a relationship is modulated in different ways under different ranges of conditions. Also, as will be seen in the Discussion, intercluster variations (differences in the mean and covariances between clusters) can be used to differentiate distinct patterns of behavior, or mechanistic interactions, between conditions.

As shown in Fig. S3, GMM analysis of LEF_{amb} , PAR_{amb} and T_{leaf} , found six distinct, compact clusters that differed in the mode of interaction among the photosynthetic and environmental parameters. Encompassing points with lower PAR_{amb} showed strong (Clusters 1,2,4 and 5) dependence of LEF_{amb} on PAR_{amb} , with little contributions from T_{leaf} . By contrast, two clusters (3 and 6), which included points at higher PAR_{amb} , showed substantial dependencies on both PAR_{amb} and T_{leaf} . These results are consistent with LEF being predominantly light-limited at low ambient PAR, but increasingly limited by temperature-dependent processes at higher PAR. The presence of these two classes of clusters indicates that PAR_{amb} and T_{leaf} are likely to affect LEF_{amb} in independent ways. The fact that the shapes of the clusters were not determined with individual slicing under the individual parameters for PAR_{amb} and T_{leaf} , but with a co-dependence on both PAR_{amb} and T_{leaf} , suggests that, under some conditions, these effects interact, e.g. T_{leaf} may affect the dependence of LEF_{amb} on PAR_{amb} .

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

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

The redox state of Q_A

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

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

reflect both primary and acclimatory factors that change (or accumulate) over different time scales. Correlations that appear in such analyses can be used to test, at least to some extent, certain models, though it is important to note that more controlled experiments will be needed to fully determine cause-effect relationships, as discussed below.

A major outcome of the experiment is that, despite the fact that measurements were made over many plants, times etc., clear patterns of responses emerged that allow us to make some broad conclusions about the responses of photosynthesis to ambient and rapidly changing light. For example, the majority of NPQ_{high} was found, in general, to be rapidly variable, suggesting that qE was the major contributor: At lower PAR_{amb} that majority of NPQ_{high} was rapidly induced (see Fig. 3C), while at higher PAR_{amb} , pre-existing NPQ was rapidly recoverable (Fig. 3E) at higher PAR_{amb} .

Another important trend was the suppression of the light potentials of both LEF (Fig. 2) and NPQ (Fig. 3) under some conditions, particularly under lower PAR_{amb} and/or T_{leaf} . Further, strong decreases in LEF_{high} were not always accompanied by compensatory increases in NPQ_{high} , implying that the productive and photoprotective light potentials can be simultaneously suppressed under certain conditions, a situation that is likely to promote the formation of reactive oxygen species and photodamage (see also below), with important implications for understanding the environmental robustness of photosynthesis [46].

Disentangling interacting environmental impacts on photosynthetic processes.

A key challenge to the field experiment approach is in teasing apart effects from different environmental factors, especially considering that such factors may be codependent or interact with each other in complex ways. For example, in visual inspection, most of parameters show apparent dependencies on both PAR_{amb} and T_{leaf} (e.g. Figs. 2-6) but, because increases in T_{leaf} are often correlated with increases in PAR_{amb} , the effects of the two parameters may have been coincidental. It may also be that the environmental parameters interacted in complex ways, e.g. high PAR_{amb} may have exacerbated the effects of low T_{leaf} . To address these issues, we applied an approach based on GMM to identify clusters representing distinct interactions among parameters. The approach is unsupervised, thus eliminating potential bias, while allowing us to test for changes in the environmental dependencies among multiple environmental parameters (Figs. S3-S12).

Analysis of GMM clusters implied that most parameters were dependent on both PAR_{amb} and T_{leaf} , and at least under some conditions these effects are independent, or that one of the two factors predominates. Thus, the effects cannot be explained simply by coincidences between increased PAR and temperatures. Moreover, the non-rectilinear shapes of the clusters suggest that the effects of PAR_{amb} and T_{leaf} were interactive, e.g., changes in T_{leaf} modulated the effects of PAR_{amb} and vice versa. Overall, these interactions are in line with well-known temperature and PAR dependence of photosynthesis, but this type of analyses can reveal the specific combination of conditions that induce distinct behaviours, allowing for assessments of the involvement of specific mechanisms (see below) and to identify genotypic or management impacts on crop resilience and productivity.

At low PAR_{amb} , we expect steady-state photosynthesis to be predominantly light-limited, and thus the effects of T_{leaf} should be low. As light increases, downstream biochemistry should become increasingly limiting. Because downstream energy storage and metabolic processes are likely to be more temperature dependent than photochemistry, this shift may allow us to distinguish between these types of limitations. Such behaviours are apparent in many of the measured parameters, e.g., LEF_{amb} , which was not substantially dependent on T_{leaf} at low PAR_{amb} , but became codependent on PAR_{amb} and T_{leaf} at higher PAR_{amb} (Figs. 2A, S3), consistent with a progressive shift from light-limitation to assimilation-limitation. Similarly, NPQ_{amb} was solely dependent on PAR_{amb} in the cluster at low PAR_{amb} , but became increasingly dependent on T_{leaf} as PAR_{amb} increased (Fig. 3A). This shift is consistent with a control of NPQ_{amb} by PAR (at low PAR_{amb}) and downstream metabolic processes, particularly at higher PAR_{amb} , e.g., due to regulation of the ATP synthase activity or cyclic electron flow (CEF) [47].

By contrast, LEF_{high} and NPQ_{high} showed much greater dependence on T_{leaf} and these differences were more pronounced when the high light was imposed on leaves at low PAR_{amb} and T_{leaf} , i.e. the opposite of what was seen for LEF_{amb} and NPQ_{amb} . Interestingly, the LEF_{high} rates achieved in leaves exposed to lower PAR_{amb} were strongly suppressed below the maximum LEF_{amb} values measured at higher PAR_{amb} (compare Figs. 2A and B). This behavior suggests that the suppression of LEF_{high} occurs when abrupt increases in light overwhelm the activation of downstream energy storage and metabolic processes. This is generally consistent with observations that the activities of metabolic enzymes are regulated

to match the availability of energy from the light reactions, which involve large suite of co-regulatory processes, as extensively reviewed elsewhere, e.g. [16,47–53], but that these responses lag behind the changes in light. The *in situ* light potential measurements afforded by MultispeQ show that these situations are very likely to occur under many field situations.

These results also imply that accurate estimates of LEF, NPQ and other photosynthetic parameters will require measurements under ambient light, because sudden changes in PAR can lead to severe perturbations in photosynthetic limitations or regulation. Attempts to “simplify” field experiments by setting PAR to some constant value will lead to strong artifacts. Such effects are vividly demonstrated by the opposite dependencies of $\Phi_{2_{amb}}$ and $\Phi_{2_{high}}$ on PAR_{amb} (Fig. 2D), and validate the use of the PAR matching feature of the MultispeQ instrument. It is important to keep in mind that the rates of acclimation may vary substantially between species, and that these may be assessed by performing more intensive experiments with variable high light and dark recovery times.

Mechanisms for controlling the light potentials of LEF and NPQ using MultispeQ field data.

The rapid reversal of NPQ_{amb} and NPQ_{high} over 10 s of dark indicated that, under our conditions, NPQ is predominantly in the form of qE (Fig. 3B and C), and thus dependent on lumen acidification and subsequent pH-dependent responses. Lumen acidification can be controlled by changes in proton influx (through changes in LEF and CEF), proton efflux through the ATP synthase and the partitioning of *pmf* into electric field ($\Delta\psi$) and ΔpH components, which in turn, are impacted by metabolic status, as proposed earlier [15,38]. Here, we explore the possible mechanistic bases for these effects, by comparing the correlations among MultispeQ measurements.

Scheme 1 illustrates three basic mechanistic models describing proposed processes that can limit the light potentials of photosynthetic and photoprotective mechanisms. The models make qualitative predictions about how the actions of each mechanistic model will impact correlations between measured photosynthetic parameters, and thus can be used as a framework for interpreting the field data introduced in Results. The expected effects on the

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

should substantially decay during the 10-s dark recovery period, resulting in increased NPQ_{high-rec}. More slowly-induced or relaxing forms of NPQ, including qI, qZ and long-lived qE, may be also present prior to and throughout the experiment. The forms should register as increases in NPQ_{rec}, but not in NPQ_{high-amb} or NPQ_{high-rec}, but given that the high light and recovery periods were only 10s long, our results do not allow us to distinguish among these possible forms.

Model 3: Photosynthetic control (PCON, Scheme 1, Model 3). PCON results from the slowing of PQH₂ oxidation at the cytochrome *b₆f* complex as the lumen becomes acidified. If PCON occurs without activation of qE, we expect a net reduction of Q_A (decreasing qL) but a net oxidation of P⁺ (increasing the 810 nm absorbance signal).

The qE and PCON models can be further subdivided [18,54]. In most cases, we expect lumen acidification accompanied by elevated *pmf*, reflected in an increased ECSt signal, which can be induced by increased proton influx into the lumen, due to increased LEF, increased CEF, or decreased conductivity of the thylakoid to protons (g_H^+) by slowing the ATP synthase. Alternatively, lumen acidification can also be associated with an increase in the fraction of *pmf* that is stored as ΔpH , by controlling the flow of counterions across the thylakoid membrane, altering the partitioning of *pmf* in ΔpH and $\Delta\psi$. In this case, acidification may occur with little or no increases in total *pmf*, or the rates of proton influx, though the current field-based data do not allow us to directly distinguish these possibilities.

These models, while not mutually exclusive, will tend to counteract each other, at least within a particular leaf. For instance, PSI acceptor side limitations will tend to inhibit electron flow, thus decreasing proton flux and *pmf* generations. On the other hand, the generation of *pmf* will tend to slow electron flow (through Models 2 or 3), thus preventing the buildup of electrons on PSI electron acceptors. However, it is important to note that, in a survey type experiment like ours, photosynthesis in different leaves may be limited by distinct processes, and thus any collection of samples may reflect various combinations of the above models.

Testing models for limitations in light potentials.

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_6f 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}^+ ($P_{700}^{+high-amb}$), 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 $qL_{high-amb}$, $P_{700}^{+high-amb}$, 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

cooperatively, depending on conditions, affecting the propensity for photosynthesis to adopt Model 2 or 3 behaviours.

The data in Figs. 8 show that, at lower T_{leaf} and PAR_{amb} , qE activation was suppressed despite light-induced increases in pmf , and that this behaviour was associated with accumulation of electrons on Q_A but oxidation of P_{700} (Fig. 8A), suggesting that, under these conditions, light-induced increases in ΔpH caused slowing of the cytochrome b_6f complex (PCON), but that the qE response lagged behind or was completely suppressed, leading to Model 3 behaviour.

It has been shown that the lumen pH-dependencies of qE and PQH_2 oxidation by the cytochrome b_6f complex are tightly coordinated, so that increased lumen acidity activates photoprotection prior to PCON, presumably to prevent the accumulation of reduced Q_A [54]. However, these experiments were performed under more slowly-changing (near steady-state) conditions in the laboratory, and our results suggest that this coordination can be defeated under real world conditions in the field, especially when T_{leaf} is low and PAR fluctuates rapidly. This discoordination can have strong implications for photodamage, as it has been shown that high thylakoid pmf can greatly accelerate PSII recombination reactions, especially when Q_A is reduced, leading to $^1\text{O}_2$ production [28,29]. It thus seems reasonable to suggest that the shift from qE to PCON at low T_{leaf} will increase the rates of photodamage.

There are several possible mechanisms by which the response of qE can be uncoupled from increased pmf . Longer-term dependencies of NPQ on temperature have been reported under both field [57–59] and laboratory [60][61] conditions. The current work shows effects on rapid NPQ and LEF changes, which can be related to distinct mechanistic models. For example, it is known that the xanthophyll cycle is strongly temperature dependent, though the general observation is that zeaxanthin tends to accumulate at lower temperatures due to a slowing of the epoxidation of zeaxanthin [60][61,62]. Interestingly, we would expect the accumulation of zeaxanthin to augment, rather than suppress qE responses as we have observed in the current results. Lumen acidification may also be rate limiting for formation of qE. While rapid increase in light can result in nearly instantaneous increases in $\Delta\psi$, formation of ΔpH and lumen acidification requires counterion transport processes, which tend to be slow, and thus lumen acidification lags behind [25,29], and it is possible that this

process is substantially slowed at low temperature. Other possible limitations include temperature-dependence of conformational rearrangement of antenna complexes following protonation of PsbS [63,64], which in turn may be related to the interactions among thylakoid proteins, lipids and ultrastructure [12,44,65,66]. The current data does not allow us to discriminate between these models with the current data set, but the work suggests conditions and species under which such limitations occur, and how they may impact plant productivity or resilience.

Conclusion: Current limitations and prospects for open science-led efforts to understand and improve photosynthesis.

There are intense, ongoing efforts to improve photosynthesis, yet the importance of the responses of photosynthesis under fluctuating, real world conditions are just now being recognised. In particular, we lack understanding of the extents and impacts of these responses, as well as their mechanisms and genomic control, which will be critical to achieving field-relevant improvements in efficiency and robustness, especially in a changing environment.

Here, we demonstrate methods and tools to assess the light responses of photosynthetic processes under real world conditions, and use them to explore the factors that limit the capacity of plants to utilise or dissipate rapidly increased PAR. A major outcome is that, despite the complexities of field environments, clear behavioural patterns can be resolved, as long as the experiment contains a sufficient number of points taken over a large environmental space, and that includes both environmental metadata. Such combinations of information allow for the generation and testing of specific hypotheses. For example, we observed no evidence for Model 1 behaviour in the current study, but we do not exclude the possibility in different species and/or different environmental conditions. The rapid measurements allowed us to test for various models over broad-scales by looking for internally-consistent relationships among the various measured parameters. Further, while we surmised (above) that Model 3 type behaviour would likely lead to photodamage, we do not have independent endpoint measurements (e.g., yield, growth rates etc.) to validate that the propensity for Model 3 behaviour has long-term consequences. The models are not exclusive, and there will almost certainly be cases, e.g., Cluster 4 in Fig. 8, where

intermediate behaviours will be apparent, either because of co-limitations among multiple processes or heterogeneity between chloroplasts in the leaf samples.

We also emphasise that the data presented here was intended to introduce the approaches and methods, and thus leaves a number of questions unanswered, but sets up the approach to further study. The origins of these effects may include several classes of processes [31,67], that may differ under different conditions [68], including induction of downstream assimilatory reactions and metabolic pools [69,70], downstream sink reactions [71], redox regulation [72,73], balancing between the production and consumption of ATP and NADPH [1,49], ion homeostasis and regulation of thylakoid *pmf* [25,74], low stomatal aperture that may lead to transient depletion of internal CO₂ levels. Distinguishing these will likely require more detailed phenotyping and biochemical [10,56,75], modeling [31] and genomics and genetics approaches [76].

The accessibility of the tools should allow larger numbers of researchers to answer these types of questions over a broader set of results, as will be presented in an upcoming paper. This approach was made possible by the combination of several open science advances. Collation of large amounts of data and metadata through the MultispeQ and PhotosynQ platforms [33], allowing us to explore the interdependencies of multiple phenotypes and environmental conditions (metadata). The GMM methods allowed us to explore the interactions among multiple environmental parameters and photosynthetic phenotypes, and test for the participation of distinct mechanistic models to explain the limitations to photosynthesis under field conditions, leading to the identification of distinct limitations in the rapid activation of NPQ and LEF at low temperature. Finally, making all tools, protocols and analytical methods available in directly usable forms, the project can be readily expanded to include multiple environments and species, as well as alternative 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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