REVIEW PAPER

Explaining pre-emptive acclimation by linking information to plant phenotype

Pedro J. Aphalo1*,2, and Victor O. Sadras2

1 Organismal and Evolutionary Biology Research Programme, Viikki Plant Science Centre, Faculty of Biological and Environmental Sciences, University of Helsinki, Finland
2 South Australian Research and Development Institute, and School of Agriculture, Food and Wine, The University of Adelaide, Australia

* Correspondence: pedro.aphalo@helsinki.fi

Received 5 December 2018; Editorial decision 30 November 2021; Accepted 6 December 2021

Abstract

We review mechanisms for pre-emptive acclimation in plants and propose a conceptual model linking developmental and evolutionary ecology with the acquisition of information through sensing of cues and signals. The idea is that plants acquire much of the information in the environment not from individual cues and signals but instead from their joint multivariate properties such as correlations. If molecular signalling has evolved to extract such information, the joint multivariate properties of the environment must be encoded in the genome, epigenome, and phenome. We contend that multivariate complexity explains why extrapolating from experiments done in artificial contexts into natural or agricultural systems almost never works for characters under complex environmental regulation: biased relationships among the state variables in both time and space create a mismatch between the evolutionary history reflected in the genotype and the artificial growing conditions in which the phenotype is expressed. Our model can generate testable hypotheses bridging levels of organization. We describe the model and its theoretical bases, and discuss its implications. We illustrate the hypotheses that can be derived from the model in two cases of pre-emptive acclimation based on correlations in the environment: the shade avoidance response and acclimation to drought.

Keywords: Adaptation, cues and signals, drought, eco-devo, epigenome, genome, information, phenome, preemptive acclimation.

Introduction

The importance of context and information in the study of plants

Current theory of the phenotype is lagging behind our fast-growing ability to generate genetic and phenotypic data (Noble, 2014; Sadras, 2021). We need conceptual models to explain and predict how these two types of data are causally interconnected, particularly for complex traits where an unjustified, unidirectional gene-to-phenotype model is implicitly still prevalent (Box 1, ‘Phenotype and downward causation’).

Context, as used in this review, includes the environments to which an individual organism and its ancestors have been exposed, and is key to understanding development, behaviour,
Box 1. Key concepts and definitions.

**Abstraction, idealization, and effective theory.** ‘An abstract description of a system leaves a lot out. But it is not intended to say things that are literally false. An idealised description of a system is a description that fictionalizes in the service of simplification…’ (Godfrey-Smith, 2009). Effective theory allows modelling of the behaviour of the system without specifying all of the underlying causes that lead to system state changes; by definition, effective theories are agnostic to system mechanics (Flack, 2017); see also coarse graining.

**Coarse graining.** Coarse graining is a reduction of the microscopic details of a system. Plants sense individual aspects of the environment such as presence of neighbours and soil drying through reasonably well-established mechanisms, but the integration of presence of neighbours and dry soil remains a gap. The Flack (2017) model of coarse graining is an interesting perspective for such integration. In this scheme, \( E \) are environmental states including presence of neighbours and dry soil, and \( P(i = \text{neighbour}) \) and \( P(i = \text{dry soil}) \) are the respective algorithms by which these components, \( i \), in \( M \) estimate environmental states. \( C \) is collective computation by \( M_i \) of \( S \), the macroscopic variable, and \( D \) is the downward causation via \( i \) in \( M \) reading \( S \) and tuning the phenotype to the integrated condition of presence of neighbours and dry soil. Contemporary efforts in quantifying collectivity (Daniels et al., 2016) could provide novel insights into plant integration of multiple cues and signals.

**Cue and signal.** Karban (2015) writes ‘…I will regard responses to stimuli as examples of plants sensing cues but not communicating.’ In the case of ‘signal’, definitions vary among authors, but in general criteria are stricter than for ‘cue’, in many cases implying communication that is beneficial to both parties involved, and that emission and sensing of the signal have evolved for the purpose of sharing information. In practice, a clear distinction between cues and signals for specific interactions is difficult (see Karban, 2015, chapter 1): signals are thought to be sent while cues happen, a distinction that in the case of plants we can only guess from the observed behaviour.

**Decision making.** We use this term as an abstraction indicating a ‘choice’ of one out of many possible development ‘paths’ available to an organism. By this, we do not imply that plants make conscious choices, or that consciousness might play a role in the model we present (see Taiz et al., 2019).

**Information.** The role of information we discuss here is related to an organism’s interaction with its environment. Our model is agnostic about considering the process of evolution itself as a flow of information or not (see Godfrey-Smith, 2013, chapter 9), which is not required to be able to consider DNA as a memory of past evolutionary events.

**Maternal effects.** Maternal effect is ‘the causal influence of the maternal genotype or phenotype on the offspring phenotype’ (Wolf and Wade, 2009), and the continuity of the phenotype refers to the ‘unbroken and overlapping connections between the generations mediated by parentally constructed offspring phenotypes (e.g. eggs, spores, seeds)’ (West-Eberhard, 2003). The seed thus carries information across generations. In organisms with no parental care, such as plants, maternal effects can be attributed to two mechanisms: offspring provisioning and epigenetics. Maternal offspring provisioning has a quantitative component, (i.e. seed mass reflecting the amount of reserves and embryo size), and the transmission of somatic or cytoplasmic factors mediated by nutrition and metabolism (Kuijper and Johnstone, 2015); some plants can also transmit microbial symbionts to the progeny, which can influence offspring fitness (Gundel et al., 2017). Epigenetics—a change in gene expression without base sequence alteration—involves processes such as DNA methylation, RNA-directed DNA methylation, nucleosome histone post-translational modifications, and regulation of small RNA activity; some of these modifications are stable and form the basis of ‘stress memory’ that is carried over across generations (Springer, 2013).

**Memory, behaviour, and problem solving.** Broadly speaking, memory is the storage of information that has been acquired through sensing of cues and/or signals. Behaviour is used in different contexts, such as psychology and mathematics, and in the second case describes the general properties of outputs given certain inputs. In this second sense is that we consider behaviour applicable to plants and the outcome of sensing of cues. Through idealization, some of this behaviour may be explained as contributing to solve a ‘problem’ faced by an organism.
growth, and reproduction. The importance of context stems from the non-additive nature of the influence of its components onto plant responses. However, context is often overlooked in the design of experiments and in the interpretation of the plant phenotype, for example when gas exchange measured in individual leaves ignores the effects of both leaf and canopy boundary layers (Jarvis and McNaughton, 1986), or when metabolic profiles of plants ignore the artefacts associated with step changes in irradiance compared with the day-night sinusoidal irradiance regime or irregular variation due to clouds (Annunziata et al., 2017), or when interference between adjacent maize plants in a greenhouse is ignored (Chen et al., 2019).

The importance of context varies. For constitutive traits, biotechnology applied to crop protection has been very successful, as illustrated in the reduced reliance on broad-spectrum insecticides for cotton and maize crops transformed to express Bt (Bacillus thuringiensis) toxins targeting lepidopteran pests (Fitt, 1994; Downes et al., 2016), and herbicide resistance in soybean favouring no-till systems (Viglizzo et al., 2011; Marinho et al., 2014). In contrast, for traits under complex regulation and naturally part of acclimation responses, biotechnology has underdelivered, as illustrated by meagre success in improving crop yield despite significant efforts (Passioura, 2006, 2020; Tardieu, 2012; Gilbert, 2016; Dalal et al., 2017a, b). Competition for resources among plants depends directly on the acquisition of resources and indirectly on the acquisition of information.
allowing prediction of future contests for resources (Ballaré et al., 1987; Novoplansky et al., 1990; Aphanl and Ballaré, 1995; Aphal et al., 1999). Thus, competitive behaviour as elicited by perception of signals and cues has temporal and rate-related constraints dependent on both a plant's stage of development and size and those of its neighbours (Novoplansky, 2009).

For crops, yield does not normally scale from single plant to stand (Pedro et al., 2012), and for natural vegetation, distribution of plant species in most cases cannot be predicted from survival of plants growing in isolation. Although neighbours are in both cases important, there are differences between wild plants and crops in their responses to them as nature selected for but agriculture selected against competitive ability (Denison, 2012; Weiner et al., 2017; Weiner, 2019; Cossani and Sadras, 2021). In addition, compared with crop stands, natural vegetation is often more diverse, leading to more complex interactions. Although context has been considered in many vegetation and ecosystem studies, our understanding of the role played by plants' multiple sensory mechanisms and informational signalling in fitness is only partial and mostly qualitative.

Many traits of ecological or agronomic relevance including fitness and grain yield result from the interaction of numerous cellular signalling pathways modulated by perceived cues and signals (Box 1, ‘Cues and signals’). For these traits, fine-tuned regulation is more important than overall metabolic capacity. Both fitness acquired through evolution and improved crop yield depend on the orchestration of the regulation of multiple developmental, morphological, physiological, and molecular characters including many not directly related to the acquisition of energy and matter (West-Eberhard, 2003).

Earlier we have argued that to understand plant–plant interactions it is not enough to consider resources because the ability of a plant to acquire these resources depends strongly on its ability to acquire and use information (Aphalo and Ballaré, 1995). This view has been supported by later research and has been influential in the development of an approach to the study of plants based on the concepts of behaviour and ‘problem solving’ (Trewavas, 2009). Twenty-five years later, here we present a conceptual model that expands the scheme of Aphalo and Ballaré (1995) by connecting the properties of the environmental context, natural selection, molecular signalling, and genetic and epigenetic mechanisms using an information-based view.

Our approach is inspired in sensory ecology and biosemiotics. Sensory ecology is a key aspect of the study of animal life (Dusenbery, 1992; Stevens, 2013). Biosemiotics, following Sharov (2016), emphasizes dynamic aspects of signs at the evolutionary and developmental time scales, featuring ‘constructivism’ in the sense that ‘…everything has to be constructed: sense organs—to detect signals; networks—to integrate and analyse signals; effector organs—to respond; memory—to store information; subagents—to perform downstream tasks including lower level construction; body—to integrate all functional units; niche—to live in; tools and resources—to increase functional efficiency; and signs—to support communication between parts of an organism and with other organisms’.

When reviewing the evolution of responses to stressors from the perspective of animal development, Badyaev (2005) stated that ‘When a stressor is reliably preceded by other environmental changes, their mutual recurrence facilitates the establishment of stressor recognition, assessment and avoidance strategies, such that an evolved stress-specific strategy does not involve an activation of an organism-wide stress response’. Novoplansky (2016) wrote that ‘[plants can] perceive, integrate and adaptively respond to myriad internal and external cues that are correlated with their future environments, in ways that maximize their life-time performance.’ Consequently, stress reactions are triggered when there is a discordance between the environment previously experienced and the current one (Badyaev, 2005); that is, when suitable anticipatory responses are not triggered in time to be effective or have not evolved. Evolution of anticipatory responses depends on the availability of suitable cues and signals. Bet-hedging is an alternative strategy that moderates risk of catastrophic failure even when the time course of environmental variation cannot be reliably anticipated (Childs et al., 2010; Shemesh and Novoplansky, 2013). Bet-hedging can, for example, be implemented as a constitutive tolerance response to a stressor that appears as a ‘waste’ of resources in the absence of the stressor.

In this review we use the terms ‘decision’, ‘memory’, and ‘behaviour’ for plants only to refer to an abstract functional role, with no reference to biological implementation and without implying volition or consciousness (Box 1, ‘Decision making and ‘Memory, behaviour, and problem solving’). As Kauffman (2016) states:

‘…E. coli must ‘sense’ its world and has done so by evolving receptors for many signals, from glucose to acidity…This sensing of its world’s possible states, as given, for example, by the bound and unbound states of receptors for glucose, hydrogen ions, and so on, constitutes ‘biosemiotics’ at its root. Once life exists, sensing of its world was of selective advantage. But given that sensing, the E. coli must ‘evaluate’ good for me and bad for me’, it must make a ‘decision’ to approach food or flee toxin, and then it must be able to act in the world to achieve an instrumental ought. Once doing exists, so do instrumental, not yet ethical ‘oughts’…”

In a theoretical analysis of the control mechanisms of annual cycles in vertebrates, Wingfield (2008) discussed the role of acclimation and fitness in variable environments. Wingfield's framework includes five categories of cues, which are relevant to account for environmental influences on the growth and yield of cereals (Sadras and Slafer 2012): (i) developmental cues (e.g. tissue interactions); (ii) initial predictive information
including environmental cues that allow long-term predictions (e.g. photoperiod); (iii) local predictive information allowing fine-tuning (e.g. rainfall or temperature); (iv) synchronizing and integrating information [e.g. social stimuli, red (R):far-red (FR) photon ratios in plant canopies]; and (v) labile perturbing factors (i.e. unpredictable environmental events).

Donaldson-Matasci et al. (2013) analysed the implications of environmental variability in cues used by organisms for predictive acclimation, and Novoplansky (2016) discussed anticipation in plants using the term ‘future perception’ to describe what we will call here biological forecasting. We prefer biological forecasting as this term better highlights the role of uncertainty in perception-based temporal extrapolation by organisms.

Resilience of ecosystems is the result of events at multiple levels of biological organization (Thorogood et al., 2020, Preprint) of which here we consider the evolution and function of anticipatory plasticity in plants. We propose a conceptual model that links developmental biology and evolutionary ecology with the acquisition of information by the sensing of cues and signals. The model is based on the idea that the plant ‘reads’ much of the information in the environment not from individual cues and signals but instead from their joint multivariate properties such as temporal and spatial correlations. Our model can be used to generate testable hypotheses at different levels of organization. Here we describe the model and its theoretical bases, and illustrate the hypotheses that can be derived from it. We apply the model to a well-understood case of pre-emptive acclimation in plants, the shade avoidance syndrome, and an additional case for which we hypothesize an information-dependent mechanism: pre-emptive acclimation to drought upon exposure of plants to UV radiation.

**Information acquisition and use**

Plants have numerous sensory systems capable of perceiving variation in the environment with high resolution (see Karban, 2015). New, unexpected senses have been described or postulated for plants, such as perception of magnetic (Ahmad et al., 2007; Maffei, 2014) and electrical fields (Hebbar and Sinha, 2002), sound (Gagliano et al., 2012) and mechanical vibration or contact (de Wit et al., 2012), and discrimination among volatile molecules or cocktails of volatile molecules (Pierik et al., 2014). Plants can communicate with each other and with other organisms using different signals (Falik et al., 2012, 2014; Pierik et al., 2014). Plants also utilize delayed responses, after-effects or ‘memory’, and spatial and temporal averaging (Sung and Amasino, 2006; Bruce et al., 2007). The capabilities for self-recognition (Gruntman and Novoplansky, 2004) and kin recognition (Crepy and Casal, 2014; Bowsher et al., 2017; Murphy et al., 2017a, b) have also been described. Kinases play a central role in perception and signalling in plants (e.g. Osakabe et al., 2013; Bourdais et al., 2015). It is noteworthy that kinases—key enzymes in cellular signalling—are more abundant in plants than in animals (Idänheimo, 2015), suggesting that metabolic signalling could, from the point of view of information processing, partly substitute for the lack of a nervous system in plants (Idänheimo, 2015). Furthermore, capacity for perception and response to signals and cues does not pre-suppose consciousness or intelligence in plants (Taiz et al., 2019).

Acclimation involves ‘decisions’ (sensu Kaufman 2016) about development, morphology, chemical composition, and physiology. Mechanistically, most often the first committed responses are changes in the expression of genes upstream of signalling cascades that can result in profound changes in metabolic pathways, plant morphology, and behaviour. For example, in the annual cycle of trees, several informational signals and their memories are a source of information for the timing of phenology and the modification of metabolism and cellular components leading to cold hardness (Hänninen and Tanino, 2011; Hänninen, 2016).

We define ‘normal acclimation’ as a response to a gradual increase of the strength of the stressor, or repeated exposure to the stressor, while we define ‘pre-emptive acclimation’ as acclimation triggered by sensing of cues or signals that usually precede exposure to the stressor. There are several well-documented examples of pre-emptive responses by plants in addition to the example in the preceding paragraph: (i) to future shading (Ballaré et al., 1987; Novoplansky et al., 1990); (ii) to changing nutrient availability in the soil (Shemesh et al., 2010, 2011); (iii) to impending drought (Falik et al., 2012; Robson et al., 2015); and (iv) to high risk of an imminent attack by herbivores (Ballaré, 2009; Karban, 2015). The complementary idea of acclimation to favourable conditions is equally true, as considering a given condition as positive or negative depends on what, we as observers, choose as the ‘normal’ reference condition, for example the photoperiodic modulation of mortality of florets in the ear of the wheat plant, whereby day length acts as a cue that anticipates the duration of grain filling (Ghiglione et al., 2008).

Acclimation of plants to stress, by definition, precedes the stress it helps tolerate or avoid. This follows from the definition of acclimation as a process that requires time and is rarely fully reversible. Within the life of an individual, its acclimation takes places concurrently with exposure to the environment, but with a lag. Fitness is determined by the dynamic interaction between genotype and environment through the life cycle (Fig. 1). This interaction involves acquisition of information by sensing cues and signals, and environmental and developmental constraints. This process repeats for each individual during each generation, driving evolution, including the evolution of pre-emptive acclimation.

Our analysis focuses on information, rather than on physiological mechanisms or ‘implementation’. This is a more abstract viewpoint, which favours generalization at the expense of mechanistic descriptions of individual cases (Box 1,
The terms constancy and contingency to name the sources of methods for time series (Colwell, 1974). Colwell (1974) used ecology and evolution, and can be analysed using statistical phenomena and their predictability play an important role in Patterns of temporal fluctuation in physical and biological environmental variation. The non-random components of the system under study. Pierik (1977) have highlighted plant’s environment needs to be included as a component of Harper, 1977). performance of whole plants in communities (Donald, 1963; Donald, 1963; Pierik and T esterink, 2014). By doing so, we will be able to capture interactions among the individual responding processes and their role in the behaviour and performance of whole plants in communities (Donald, 1963; Harper, 1977).

When we ask questions related to fitness and evolution, the plant’s environment needs to be included as a component of the system under study. Pierik et al. (2014) have highlighted the need to take into account the community in which the plants grow; here we add the abiotic environment and, most importantly, the statistical relationships among the various biotic and abiotic variables. However, as Stevens (2013) emphasized for animals, we should do this with reference to the sensory abilities of each species. Most research of plants’ sensory capabilities has centred on the plant and its responses rather than on describing the multivariate dynamics of the plants’ environment. Existing studies are few and frequently limited to the dynamics of aggregate summary variables (e.g. R:FR photon ratio versus herbaceous canopy development, Evers et al., 2006), or long-term dynamics (e.g. species succession and seasons in forests, Ross et al., 1986).

Fig. 1. Pre-emptive acclimation and selection: orange, time course of one realization of the environment (E) during the lifetime of an individual of a genotype (G); teal, time course of phenotype (P) through development, growth, and acclimation; black, sensing of cues and signals targeted and initiated by the plant, leading to acquisition of information; red, selective pressure from the environment; green, (time-consuming) acclimation response. The phenotype is the outcome of the expression of the genotype in an environment, P(G, E, G×E), where G×E describes the non-additive interaction. Filled arrowheads indicate direct dependence on the environment, while open arrowheads indicate dependence mediated by the genotype and phenotype. For simplicity, we plot continuous time as discrete steps.

‘Abstraction and effective theory’). The difference between the usual metabolic signalling diagrams and an information-based model is that the abstractions are based on different criteria, suitable for the analysis of different types of questions: those related to proximal mechanisms and those enquiring about ultimate evolutionary causes.

In addition, when studying acclimation and adaptation, we are concerned with the performance of whole plants. Consequently, even when dealing with mechanism, it is best to study responses as syndromes affecting whole individuals rather than responses of isolated processes or features (Aphalo, 2010; Pierik and T esterink, 2014). By doing so, we will be able to capture interactions among the individual responding processes and their role in the behaviour and performance of whole plants in communities (Donald, 1963; Harper, 1977).

When we ask questions related to fitness and evolution, the plant’s environment needs to be included as a component of the system under study. Pierik et al. (2014) have highlighted the need to take into account the community in which the plants grow; here we add the abiotic environment and, most importantly, the statistical relationships among the various biotic and abiotic variables. However, as Stevens (2013) emphasized for animals, we should do this with reference to the sensory abilities of each species. Most research of plants’ sensory capabilities has centred on the plant and its responses rather than on describing the multivariate dynamics of the plants’ environment. Existing studies are few and frequently limited to the dynamics of aggregate summary variables (e.g. R:FR photon ratio versus herbaceous canopy development, Evers et al., 2006), or long-term dynamics (e.g. species succession and seasons in forests, Ross et al., 1986).

The non-random components of environmental variation

Patterns of temporal fluctuation in physical and biological phenomena and their predictability play an important role in ecology and evolution, and can be analysed using statistical methods for time series (Colwell, 1974). Colwell (1974) used the terms constancy and contingency to name the sources of predictability. Since the 1970s, the analysis of time series has developed, extending its scope to include multivariate data as well as discrete events. The idea that temporal variation can be assigned to different generating mechanisms or processes and that these processes can contribute to predictability, remains valid.

To a large extent, variation in the environment has structure: variables do not vary independently of each other, or independently of their previous or future states. Hence, current and past states of variables can be a source of information for prediction of the future state of the same variable, the current state of different variables, or the future state of other variables. For any organism, predicting future conditions can be expected to contribute to fitness. Conditions include both normal events, which occur frequently, and infrequent extreme events—once over many generations. These uncommon events can impose limits to evolution (Gutschick and BassiriRad, 2003; Lyberger et al., 2021).

From this it follows that within the constraints of the evolutionary process, and the reliability of available sources of information, most organisms, including plants, should be expected to acquire, store, process, and use information during their lifetime in decision making (Box 1, ‘Decision making’) related to acclimation. We should be aware, though, that predictability of events creates boundaries to the plastic behaviours that can persist in the long run versus bet-hedging strategies (e.g. Childs et al., 2010; Grantham et al., 2016). Natural selection of survivors to exceptional events may lead to behaviour that can be described as ‘risk aversion’ (Novoplansky, 2009).

Describing correlations and lags among environmental variables is crucial for understanding their role as sources of information for pre-emptive responses that depend on implicitly ‘forecasting’ future events. Autocorrelation describes correlation in time for a variable with itself; it is typical of gradual, cyclical, or repeating patterns of change. Cross-correlations describe the ‘parallel’ changes of two or more variables in time. If there is lag in a cross-correlation, it means that variation in one variable is consistently delayed compared with the variation in another variable, while both variables follow a similar pattern of temporal change.

Some patterns of variation are both cyclic and deterministic, such as day length. In such a case, the future state of the
variable can be predicted if the period, amplitude, and phase are known (see Fig. 2A for a simple example). Two such patterns can be shifted in time, and the early one directly used to predict the future state of the later one (Fig. 2B). Many patterns of environmental variation are not fully deterministic, but nonetheless are not completely random because of the presence of correlations. The simplest case for a time series is autocorrelation, in which values close in time are more similar than those further away in time. This kind of pattern can be simulated using random variation as a starting point (Fig. 2C). This demonstrates that information about the correlation acting on a random process is useful for forecasting the future state of a variable using its current or recent state as input.

In nature, these components jointly contribute to the observable variation such as cyclic, and random autoregressive (Fig. 2D). These latter examples are presented for a single variable for simplicity, but correlation among ‘noisy’ variables can also provide useful information for the prediction of the future state of lagged variables. Until now, we have centred the discussion on changes in the time domain. Similar correlations exist in the spatial domain. In certain cases, lag in time is caused by differences in the speed of propagation in space. The temporal lag between two signals originating at the same point in space, but propagating at different speeds, depends on the distance travelled and their relative speeds.

It is important to realize that when such lags or correlations among variables are not part of the physical and chemical environment, organisms have the ability to ‘add’ signals to their environment that do have these properties. For example, the emission of plant volatile organic compounds (VOCs) in response to herbivory could generate a signal that propagates faster to neighbouring plants than the insects move, resulting in a delayed arrival of the insects with respect to the arrival of the VOC signal. In addition, as the activity of the herbivores triggers the emission of VOCs, the presence of VOCs in the air in the neighbourhood of a plant under attack is tightly correlated with the (impending) arrival of the herbivores. It must be stressed that here we are discussing correlations, and consequently the previous statement should be interpreted as the probability of insects soon reaching the target plant being higher when VOCs are present in the air than when they are not.

Superimposed on environmental patterns there is a significant amount of ‘random noise’ or variation to which we are unable to assign a deterministic origin. Statistics gives us the tools, as researchers, for separating interesting information from random variation or, so-called, noise (Box 1, ‘Noise’). Statistical algorithms can be computed in analogue systems as well as in digital ones, and it has been proposed that even primitive

---

**Fig. 2.** Artificial examples of patterns of environmental variation (t depicts time, and y the value of an arbitrary environmental variable). (A) Deterministic cyclic variation; (B) the same as (A) (in black), but adding a second variable with the same amplitude and cycle but lagged (in red); (C) an autocorrelated time series generated from a random process; (D) a combination of cyclic deterministic variation and autocorrelated ‘noise’ in the response, AR(y). See Box 2 for calculation details.
organisms can do ‘maths’ through metabolic signalling (Daniel et al., 2013). As in the case of statistical time series analysis, different sampling and smoothing methods can be expected to play a role in information processing by organisms. Even sharing of information among neighbours may, in some cases, be equivalent to sampling and averaging over a larger area, which could be beneficial to all plants involved in the case of variables with dynamic spatial heterogeneity in their state, such as herbivory.

The needed ‘information processing’ can also be complex in the time domain because the timing of a response can be crucial for fitness. A cue such as night length is minimally affected by noise (Box 1, ‘Noise’) and consequently a very reliable source of information; even though night length is a reliable cue, its correspondence to seasons of the year is not monotonic:

In Fig. 2 we show plots of time series artificially generated in R (R Core Team, 2021) assuming different generative processes. We describe here the algorithms used to generate each of the time series accompanied by brief explanations.

**Figure 2A** and **B** are the result of deterministic processes with cyclic variation with no random component. Based on arbitrary $t$ values, representing an ordered sequence of distances in time or space from an origin, $y$ values were computed without a lag as

$$y_i = f(t_i),$$

and with a lag as

$$y_i = f(t_i + l),$$

where $f$ is a determinist cyclic function such as sin or cos and $l$ is a lag (i.e. a constant shift along the $t$ axis).

In the remaining panels, we use as a starting point a series of (pseudo)random values generated from the Normal distribution $N(\mu, \sigma)$. The series in **Fig. 2C** has no deterministic component, but it is generated by an autoregressive, AR($y$), process where the value of $y$ at the next time step $t_{i+1}$ depends on a random component and the value of $y$ at time $t_i$. The series is generated recursively advancing one step of 8 at a time using R function `diffinv()` applied to a vector of independent and normally distributed values

$$y_{i+1} = y_i + N(\mu = 0, \sigma = 1).$$

The series in **Fig. 2D** combines the deterministic cyclic component from **Fig. 2A** and the autoregressive random component of **Fig. 2C**.

---

**Box 2. Correlations in the environment.**

Environmental variables are not independent and identically distributed (iid).

The state of individual variables is autocorrelated both in time and in space; for example, a warm day is more likely to be followed by another warm day than by a much colder day. The same is also true spatially, the soil water content 0.1 m away from the current location is more likely to be similar than that 1.0 m away. From a multivariate perspective, different environmental variables are correlated with each other; for example, within a single day, water vapour pressure in the air tends to vary little, but near noon when air temperature is higher, the vapour pressure deficit is usually at its maximum and relative humidity at its minimum. The mechanistic explanation behind these different correlations varies, but irrespective of their origin correlations carry information useful in forecasting. Information we also intuitively use in everyday life.

The needed ‘information processing’ can also be complex in the time domain because the timing of a response can be crucial for fitness. A cue such as night length is minimally affected by noise (Box 1, ‘Noise’) and consequently a very reliable source of information; even though night length is a reliable cue, its correspondence to seasons of the year is not monotonic:

Each night length occurs twice per year in opposite seasons. In contrast, daily temperature is affected by strong variation in its temporal course, with patterns changing year to year due to prevailing weather conditions. These differences in the quality of the information source lead to different strategies in its use. For reproductive induction by short nights, a single short night event can inform about seasonal timing—leading to experimental observations of a single night break inducing flowering in some species (Jackson and Thomas, 1999). In contrast, temperature requirements for developmental events are most frequently a combination of previous ‘accumulated’ low or high temperatures and current temperatures (Baulcombe and Dean, 2014). An example of the use of multiple cues, functioning on a shorter time scale, is the complex interplay of cues perceived through different photoreceptors (Casal, 2013; Rai et al.,
that also includes the temporal integration of these cues through the day, which apparently prevents a premature, or too strong, shade avoidance response under moderate shade (Casal, 2012; Sellaro et al., 2012). In other cases, redundant sources of information can substitute for each other: for seed germination in many species, the well-known ability of alternating day/night temperatures to substitute for or modulate a high R:FR photon ratio requirement can be thought as having a partly overlapping role in the detection of bare (unshaded) and own−depth-in-the-soil for seeds (Benech Arnold et al., 1988; Vazquez-Yanes and Orozco-Segovia, 1994). This redundancy, possibly stemming from the dual role of phytochromes as light and temperature sensors (Casal and Balasubramanian, 2019), can be thought as reflecting an overlap in information content between two environmental cues. Both qualitative and quantitative cues may provide information, but the adaptive advantage of responses to cues depends on the local environment as a whole, leading to broad genetic variability in natural populations (see Murfet, 1977). For example, in addition to the well−known correlation of photoperiodic responses of plants to seasonal variation in temperature, similar correlations to the local timing of the rainy season have been described (Murfet, 1977; Ryan et al., 2016). The more and better information is available—including on the context—the more reliable forecasts tend to be (Hyndman and Athanasopoulos, 2018). What we know about plants indicates that the regulation of metabolism and development relies on multiple sources of information combined through complex signalling networks containing multiple feedback loops and points of interaction (Ballaré and Pierik, 2017; Rai et al., 2021). This is at the core of why extrapolating the results of experiments done in an artificial context into natural or agricultural systems almost never works for characters whose environmental regulation is important for the organism’s fitness: biased relationships among the states of different variables in both time and space may disturb the information decoded by the plant, returning ‘accidental’ phenotypes (e.g. Annunziata et al., 2017) due to a mismatch between the selection history reflected in the genotype and the artificial growing conditions guiding its expression into phenotype. In addition, at the metabolic and signalling level, organisms have redundant paths for regulation, and compensatory regulation may mask the effect of altering one or few components (Ovaska et al., 1992; West−Eberhard, 2003; Noble, 2012). Inconsistent results under controlled and natural environments are common, and are a bottleneck for the directional biotech pipeline from lab to field (Chan et al., 2020). One striking example is that of the UV-B photoreceptor UVR8 in Arabidopsis: UVR8 dysfunction was reported as highly detrimental to growth in a unique sun simulator chamber designed to simulate the natural radiation environment (Favory et al., 2009). However, that uvr8 mutants can survive and flower in sunlight (Morales et al., 2013) and grow normally in the same growth chamber under a slightly different illumination regime suggests that small differences in the timing of UV-B exposure within the photoperiod are important for tolerance (Rai et al., 2019). Only considering the spectral properties of sunlight together with the spectral and photochemical properties of the UVR8 photoreceptor has allowed an understanding of how plants perceive solar UV radiation (Rai et al., 2021).

Life history, development, allocation, morphology, and physiology adapt and acclimate in coordination, and in the case of crops contribute to yield. For example, theoretically it should be possible to improve the energy conversion efficiency of the C3 metabolism in plants (reviewed by Raines, 2011; Evans, 2013; Furbank et al., 2015; Reynolds et al., 2021). However, a lack of understanding of how and why such apparent inefficiencies may contribute to overall plant fitness makes setting physiological targets for crop breeding extremely difficult (Denison, 2015). The complexity of metabolic interactions, trade−offs between traits, issues of scale and levels of organization, and environmental factors over−riding genetic variation converge to constrain the opportunities for breeding and selection for higher photosynthesis (Denison, 2012; Sadras and Richards, 2014; Furbank et al., 2015; Sinclair et al., 2019). Similarly, genetic modification targeting improved drought tolerance in crops has rarely been successful (but see González et al., 2019), while traditional breeding has allowed a sustained improvement of yield in dry environments for many decades (Sadas and Richards, 2014; Passioura, 2020).

The current poor record of success does not mean that indirect, trait− or genetics−based, attempts at crop improvement are inherently of little use. Instead, it shows that the dominant conceptual model of crop phenotype has been misconstrued or oversimplified; it has, among other things, failed to account for traits related to acclimation, which depend on signalling networks and coordination of multiple responses that capture the complexity of environmental variation.

**Strategies**

According to DeWitt and Langerhans (2004), plants have evolved four contrasting strategies in response to environmental variation: (i) specialization, whereby a single phenotype is produced that is well adapted to a particular environment even though the specialist may experience a range of environments; (ii) generalization, whereby a ‘general purpose’ phenotype is produced, with moderate fitness in most environments; (iii) bet−hedge, whereby an organism produces either several phenotypes (e.g. among units in a modular plant, such as co−existence of juvenile and adult leaves in some trees) or single phenotypes probabilistically (e.g. morphological and dormancy polymorphism in seeds produced by an individual plant); and (iv) phenotypic plasticity, whereby alternative phenotypes are produced in response to environmental cues. Modelling these four strategies under the assumption of perfect phenotypic plasticity and a simplified range of environments returned a ratio
of fitness after four generations of 1:1.6:1.5:25 (DeWitt and Langerhans, 2004). The conclusions from this type of analysis are that in the absence of constraints, unrestricted plasticity is superior in variable environments, and the fact that unrestricted plasticity is not ubiquitous suggests the existence of ubiquitous constraints. The more likely constraints include a relatively high cost for plasticity, developmental constraints, and unreliability of environmental cues that guide development (DeWitt and Langerhans, 2004; Sadras and Slafer, 2012; Murren et al., 2015).

A given phenotype can follow different strategies in relation to different features of its environment and, in addition, the degree of phenotypic plasticity can concurrently differ between plant traits. A genotype may express a trait that is very responsive to environmental cues, such as internode elongation versus R:FR photon ratio, but less responsive to other traits. Although the degree of plasticity is trait dependent, evidence supports partial rather than full independence between the genetics of a trait (e.g. phenology or grain weight) and the genetics of the trait’s plastic response to the environment (Reymond et al., 2003; Lacaze et al., 2009; Marguerit et al., 2012; Álvarez-Prado et al., 2014; Sadras et al., 2016), as anticipated by Bradshaw in the 1960s (Bradshaw, 1965). An important consequence of the partial independence in the genetic control of plasticity and the trait per se is that plasticity can evolve independently of the trait (David et al., 2004; Pigliucci, 2005; King and Roff, 2010). Novoplansky (2009) discussed the implications of plasticity itself being plastic, using the term metaplasticity while emphasizing risk management and plant–plant interactions.

A less frequently discussed aspect of these strategies is that many morphological and developmental responses of plants are slow compared with the speed of change in availability of resources. Moreover, such responses depend on the use of photosynthates, mineral nutrients, and other resources of limited availability. Consequently ‘valuable’ resources need to be invested, which may be recovered for re-use only at a very significant ‘loss’ (Bloom et al., 1985). For example, benefits to plants from responding to current light quality cues may depend on forecasting, or anticipating, how much and how fast neighbours will grow (Novoplansky, 1991).

**Decision making**

The use of economic models as an analogy for describing regulation of metabolism, capture, and allocation of resources has a long tradition in biology in general (Ghiselin, 2000) and plant ecology (Bloom et al., 1985). Here we highlight a specific aspect of this analogy, which has not been previously used in plant research: the analogy between the use of information and forecasting tools in dynamic resource allocation in human enterprises and the equivalent dynamic regulation of investment of limited resources by plants. Keeping this analogy in mind while reading the rest of this review is important for understanding the logic behind our conceptual model.

Acclimation, as a form of investment, can be based on continuous dynamic adjustment of allocation, for example growth allocation to shoots versus roots, or on a switch-like choice of a developmental programme, such as a switch from vegetative to reproductive stage. Reality has more nuances but, as a working classification, acclimation and development decisions can be considered as discrete alternatives or the value on a continuous scale used as set points of a feedback or feed-forward control mechanism. West-Eberhard (2003) defines a switch point as ‘a point in time when some element of the phenotype changes from a default state, action, or pathway to an alternative one—it is activated, deactivated, altered, or moved.

Even if there are recognizable patterns, the stochastic component of the environment (Fig. 2) means that ‘acclimation-related decisions’ cannot be hard-wired. These decisions need to be taken ‘on-the-go’ during plant development and are subject to errors. This brings in the interplay of profit and risk. Different contexts, and different variables within a given context, will be subject to different amounts and types of variation. From the point of view of evolution, optimization of individual traits such as carbon acquisition or the use of water during photosynthesis cannot be thought as the ‘end target’ of natural selection or the best target for crop breeding (Sadras and Denison, 2016). We should expect risk avoidance to play a key role in long-term selection (Novoplansky, 2009). As plant species differ widely in their reproductive strategies and life histories, mechanisms for risk avoidance can also vary widely. For example, in plant species propagating mainly through seeds, completion of the life cycle, and successful reproduction in every generation could be thought of as mandatory for fitness (Amir and Cohen, 1990). However, mechanisms such as the maintenance of a large and long-lived seed bank in the soil can play the role of a ‘safety net’, allowing the survival of a population and its rapid recovery after exceptional catastrophic events.

As mentioned above, in some cases such as seed germination, decision making consists of a choice between discrete options, in this case binomial: to initiate growth of the individual as a whole or not. In other cases, it can be thought of as the adjustment of a set point on a continuous scale, for example the shoot:root ratio or the regulation of stomatal conductance. In this last example, it can even be thought of as a decision to change responsiveness. For example, long-term exposure to UV radiation can trigger a reduction in the response speed of stomata to a step increase in irradiance of specific colours of visible light (Aasamaa and Aphalo, 2016, 2017).

A parallel exists between these ideas and the management of limited resources by human enterprises. Decision makers use forecasting tools, based on statistics, in particular time series analysis, combined with information about the current market and economic situation to improve the long-term return from limited resources. One successful example is the management by power utilities of power generation and distribution capacity based on demand forecasting (Hyndman and Athanasopoulos,
This parallel extends to other kinds of predictions (see Orrell, 2006; Kauffman, 2008), but we here emphasize the parallel between how organisms can achieve pre-emptive acclimation and statistical forecasting methods. If approached from a high level of abstraction, it can be seen that equivalent information sources and tools are used by human forecasters and organisms. The complex statistical models stored as computer programs and used for forecasting electricity demand in the above example, are equivalent to signalling networks and sensory mechanisms in an organism’s genome and used to ‘make favourable decisions’ on the use of limited resources frequently enough to allow both short-term fitness and long-term survival. The parameterized instances of these models could be thought of as equivalent to the genotype as expressed in different phenotypes.

Another parallel between the use of forecasting for resource allocation by human enterprises and organisms is that in both cases the context or environment is under directional change, for example technological progress and the availability of raw materials for economic markets versus evolution of other species and global change for organisms. This means that the criteria and models used in decision making need to evolve, and their performance will also depend on the decision making by the rest of the community of managers as well as by other organisms in a biological community.

A further parallel, exceeding the scope of the present paper, is that consistency of decision criteria—embedded in similar predictive models—used by different traders and enhanced by reflexivity can exacerbate the risk of widespread financial losses (Beunza and Stark, 2012) while consistent responses among neighbouring plants can lead to excessive competition and even population collapse (i.e. in the absence of clear winners and losers, e.g. Yastrebov, 1996). Competition is detrimental to yield in crop stands of homogeneous genotypes with strong photomorphogenic responses (Boccalandro et al., 2003; Pereira et al., 2017; Wies and Maddonni, 2020). In both cases the combined behaviour of players driven by positive feedback—called resonance in Beunza and Stark’s text—can result in decisions that are bad for all players both individually and collectively, providing a further example of the importance of context.

These parallels allow us to borrow concepts and approaches used in statistical forecasting and to apply them to the development of a conceptual model for the functioning and evolution of pre-emptive acclimation in plants.

**Model**

The model we present describes the use of information by organisms as a means of ‘deciding’ when and how to pre-emptively acclimate. If acclimation takes place before an organism is exposed to an event itself, either favourable or stressful, and this acclimation is triggered frequently before the actual event occurs, but only rarely when it does not occur, we can conclude that the organism has been able to forecast the occurrence of the event with a certain degree of success—with success defined as a pre-emptive response that increases fitness. As explained in the previous section, the parallel with statistical forecasting holds in many respects. In statistical forecasting, one possible approach is to use long-term time series data to develop a mathematical model, which is used together with recent and current data to forecast the future evolution of the demand for, for example, electrical power. Our model assumes a similar scheme for organisms, with the genome (viewed as a template for alternative development paths and behaviours) as the equivalent of the mathematical model of the data analyst, and the organisms’ sensory mechanisms and short-term memory as the equivalent of the short-term data acquisition and processing used by analysts in decision making (Fig. 3).

Our model is set at a high level of abstraction (Box 1, ‘Abstraction, idealization, and effective theory’) and provides a further example of the importance of context.
the basis for theory. It considers information acquisition, storage, and use, without consideration of perception, transmission, storage, and processing mechanisms. It is an idealization in that we focus on information storage, flow, and use, and only consider acclimation to a single kind of future stress or favourable situation at a time. The novelty of our model is in explicitly taking into account simultaneously several possible sources of information and their joint statistical properties as inputs for decision making leading to pre-emptive acclimation in organisms.

We define three types of storage of information: genome, epigenome, and phenotype, which span from evolutionary to intragenerational time scales. The mapping of these three stores of information onto chronological time thus depends on the life history of the organism.

We need to distinguish between maternal effects broadly understood and epigenetic regulation (Box 1, ‘Maternal effects’). The second is clearly a regulatory step involving mainly, if not only, information. We consider maternal effects dependent on resources (offspring provisioning), such as those associated with seed nutrient content or seed size, as part of the phenotype. This distinction is coherent with the use of information as an abstraction.

The model assumes that as a consequence of natural selection, the use of different cues for acclimation is not necessarily related to cause and effect relationships in the environment. As long as a correlation exists that allows the organism to forecast a future event, evolution will favour the use of this cue as a source of information. From a statistical viewpoint, evolution generates a template for pre-emptive acclimation comparable with an empirical statistical forecasting model.

An important corollary is that the overall contribution of pre-emptive acclimation to fitness is not deterministic. Pre-emptive acclimation is a risk-taking game based on the probabilities and frequencies of occurrence of different events and the quantitative benefits and drawbacks from alternative patterns of capture and allocation of resources. All this is working within the boundary set by a probabilistic risk of population extinction—a binary response.

Our model integrates environmental factors to the extent that they are structured as described above. Further integration is beyond the scope of this paper, but coarse graining can be added in future versions (Box 1, ‘Coarse graining’).

Example cases

To demonstrate the usefulness of our conceptual model for understanding the evolution of pre-emptive acclimation in plants, we will now describe two cases. One of them is the well-understood syndrome of shade avoidance, and the other is the poorly understood and controversial pre-emptive acclimation to drought mediated by plants’ exposure to solar UV radiation.

Shade avoidance and pre-emptive acclimation

Shade represents for plants a restriction on the available photosynthetically active radiation (PAR), and in vegetation canopies shade is caused by neighbouring plants. The predominant strategy of sun-adapted plant species is to reduce this shading by increasing stem length and decreasing ramification, namely a shade avoidance syndrome (SAS). In plant canopies, low R:FR ratios are correlated with the presence of neighbouring plants that are alive (Smith, 1981), consequently plants can use the R:FR ratio as a source of information on the presence, size, and distance to neighbours. Furthermore, because FR radiation is not only transmitted but also reflected by plant leaves, the change in R:FR ratio starts well before any depletion in PAR. This time offset allows the triggering of the SAS before actual shading and contest for resources starts (Ballaré et al., 1987).

The ecology of responses to neighbours and shade mediated by perception of changes in spectral composition and irradiance was thought to be well understood after a long period of study (Holmes and Smith, 1977a, b; Smith, 1981; Deregibus et al., 1983; Ballaré et al., 1987); however, significant recent progress in understanding the physiological and molecular mechanisms (Casal, 2013) has been linked to identification of new ecological functions. Several recent publications have brought to light new and exciting details showing that plants are able to use much more than the R:FR photon ratio and blue irradiance as sources of information (Casal, 2013). Perception of UV radiation is also involved in acclimation to shade (Casal, 2013; Hayes et al., 2014; Aasamaa and Aphalo, 2016; Moriconi et al., 2018). A response to the blue:green photon ratio has been described as an additional cue of shade (Sellaro et al., 2010). The same cues elicit different responses if received at different times of the day (Sellaro et al., 2012), and temporal variability (i.e. sunflecks) affects responses compared with constant illumination (Sellaro et al., 2011). Ethylene may be either a signal or a cue of the presence of neighbours in some environments (Pierik and de Wit, 2014). Physical contact could play a role when neighbours are growing very close together (Pierik and de Wit, 2014). The integration of the different cues is complex, and we lack an understanding of how the perception of neighbours works as an integrated whole (Pierik et al., 2014; de Wit et al., 2016; Ballaré and Pierik, 2017). In Fig. 4, the proposed model is applied to the flow of information involved in pre-emptive acclimation to shade. As we have considered together multiple cues of impending shade and ignored constraints, the model is a drastic simplification of reality. However, it allows us to derive useful testable hypotheses; for instance, (i) that light quality cues will trigger shade avoidance responses and (ii) that maternal effects on the readiness to respond to these cues will be relevant in species where seed dispersion is restricted to the neighbourhood of mother plants—when offspring are likely to grow in a very similar environment to mother plants.
Explaining pre-emptive acclimation

Water availability is a major driver of ecosystem structure and function, regional patterns of land use, and global agricultural productivity (Ryan et al., 2009; Chapin et al., 2011; Stewart and Lal, 2018), hence the widespread interest in plant adaptation to drought (Morison et al., 2008; Reynolds and Tuberosa, 2008; Kadam et al., 2014). In the words of Tardieu (2012) ‘any trait or trait-related allele can confer drought tolerance: just design the right drought scenario’. This highlights the importance of context once again: tailoring adaptive traits to specific environments requires quantification of natural spatial, probabilistic drought patterns in terms of timing, intensity, and duration of water stress (Chenu, 2015). Going a step further, as discussed above, various cues and signals could function as sources of information for pre-emptive acclimation, adding further constraints to realistic drought scenarios. It has been shown that plant roots can perceive local soil drying before it affects the water status of a plant (Tardieu et al., 1992; Wilkinson and Davies, 2010). This informs on the supply side of the water budget in relation to the soil volume already explored by the roots. The demand side of the water budget is described by evapotranspiration (ET), which for vegetation depends on reference or potential ET (ET0) and soil moisture (Monteith and Unsworth, 2008). In the absence of new precipitation, cumulative ET will determine the amount of water remaining in the soil at a future point in time.

In this context, we ask how pre-emptive acclimation could help to improve fitness of wild plants and yield of crops under dry conditions. In this section, we use our generic model (Fig. 3) to describe a possible mechanism for the triggering of pre-emptive acclimation to drought by exposure to UV radiation (Fig. 5). We derive testable hypotheses and demonstrate using preliminary data how these hypotheses can be investigated. Before presenting the model, we justify why solar UV radiation is worthy of consideration in a context of multivariate correlations.

The interaction between UV-B exposure and drought tolerance, for plants growing outdoors, was first described in the

Soil drought and pre-emptive acclimation

Water availability is a major driver of ecosystem structure and function, regional patterns of land use, and global agricultural productivity (Ryan et al., 2009; Chapin et al., 2011; Stewart and Lal, 2018), hence the widespread interest in plant adaptation to drought (Morison et al., 2008; Reynolds and Tuberosa, 2008; Kadam et al., 2014). In the words of Tardieu (2012) ‘any trait or trait-related allele can confer drought tolerance: just design the right drought scenario’. This highlights the importance of context once again: tailoring adaptive traits to specific environments requires quantification of natural spatial, probabilistic drought patterns in terms of timing, intensity, and duration of water stress (Chenu, 2015). Going a step further, as discussed above, various cues and signals could function as sources of information for pre-emptive acclimation, adding further constraints to realistic drought scenarios. It has been shown that plant roots can perceive local soil drying before it affects the water status of a plant (Tardieu et al., 1992; Wilkinson and Davies, 2010). This informs on the supply side of the water budget in relation to the soil volume already explored by the roots. The demand side of the water budget is described by evapotranspiration (ET), which for vegetation depends on reference or potential ET (ET0) and soil moisture (Monteith and Unsworth, 2008). In the absence of new precipitation, cumulative ET will determine the amount of water remaining in the soil at a future point in time.

In this context, we ask how pre-emptive acclimation could help to improve fitness of wild plants and yield of crops under dry conditions. In this section, we use our generic model (Fig. 3) to describe a possible mechanism for the triggering of pre-emptive acclimation to drought by exposure to UV radiation (Fig. 5). We derive testable hypotheses and demonstrate using preliminary data how these hypotheses can be investigated. Before presenting the model, we justify why solar UV radiation is worthy of consideration in a context of multivariate correlations.

The interaction between UV-B exposure and drought tolerance, for plants growing outdoors, was first described in the

**Fig. 4.** Flow of information in pre-emptive acclimation to shade by perception of radiation changes. Arrows represent flows of information: blue, retrieved from the genome (stored during earlier generations); black, acquired and/or ‘memorized’ during an individual’s or its progenitor’s lifetime; teal, regulation of gene expression by phenome or downward causation; red, lagged correlation between early changes in spectral irradiance and future low PAR irradiance; orange, outcome of information processing: a ‘decision’, based on an ‘implicit forecast of impending shade’, leading to developmental adjustments that would increase the probability of higher fitness in the presence of neighbours in comparison with phenotypes lacking pre-emptive acclimation; green, ‘shading mitigated’ compared, in probabilistic terms, with no acclimation. Dashed boxes and arrows represent the likely or forecasted future.

**Fig. 5.** Information flow in pre-emptive acclimation to drought by perception of UV-B radiation and soil moisture. Arrows represent flows of information: blue, retrieved from the genome (stored during evolution); black, acquired during an individual’s or its progenitor’s lifetime; teal, regulation of gene expression by phenome or downward causation; red, lagged correlation between UV-B radiation and drought (e.g. low soil water content and high evaporative demand); orange, outcome of information processing: a ‘decision’, based on an ‘implicit forecast of impending drought’, leading to developmental adjustments that would increase the probability of higher fitness under drought in comparison with phenotypes with no pre-emptive acclimation; green, ‘drought tolerated’ compared, in probabilistic terms, with no acclimation. Dashed boxes and arrows represent the likely or forecasted future.
context of stratospheric ozone depletion (Petropoulou et al., 1995). Gitz and Liu–Gitz (2003) concluded that UV-B radiation could enhance drought tolerance in plants through photomorphogenic effects such as decreased leaf area, but added the caveat that drought tolerance could also result from strategies other than limiting water loss. More importantly, they highlighted the need to study the effect of UV-B exposure on the tolerance of drought stress by applying these treatments sequentially instead of concurrently as had been usual until then.

More generally, it has been suggested that perception of UV-B radiation through the UVR8 photoreceptor contributes to protection from various stressors (Hideg et al., 2013; Singh et al., 2014). In sunlight, because of the shape of the solar spectrum, UVR8 mediates the perception of both UV-B and UV-A2 radiation, namely solar radiation of wavelengths shorter than ~340 nm (Rai et al., 2021). In an experiment comparing filters transmitting and attenuating solar UV radiation, we observed a strong effect, with near-ambient UV-B exposure preceding drought drastically enhancing drought tolerance. UV exposure as perceived by plants and future water availability to inform about future drought. (ii) Responses triggered by UV-B+UV-A2 radiation will enhance future tolerance of drought through signalling mechanisms that can be traced to the perception of the cue. (iii) If UV-B and/or UV-A2 radiation function as a purely informational cue, rather than as a stressor, this cue must be perceived through a photoreceptor.

To test hypothesis (i), which entails multivariate aspects of the environment, we looked for correlations between ET0 and different wavebands of sunlight using observations with very high temporal resolution for two growing seasons (Aphalo and Sadras, 2021). All bands of the solar spectrum when measured above the canopy are good predictors of ET0, including UV-B and UV-A radiation (Aphalo and Sadras, 2021). UV-A and UV-B radiation perform best at predicting variation within the photoperiod (Fig. 7), and longer wavelengths at predicting day to day variation in ET0 (Fig. 6). That solar irradiance and its components are good predictors agrees with the central role of the energy balance in evaporative demand and ET0 (Penman, 1948; Aphalo and Sadras, 2021). We concluded that UV-B exposure is an environmental cue carrying information useful for assessing the force driving ET. However, other regions of the solar spectrum carry similar information. Vapour pressure deficit and UV-B irradiance are also correlated within the course of the photoperiod (Aphalo and Sadras, 2021) as UV-B irradiance increases more with solar elevation than longer wavelengths. Even though the relationship between UV-B irradiance and evaporative demand is curvilinear, it can provide information about the demand side of the soil water balance equation. The relationship between actual ET and solar UV-B irradiance, and its consequences for soil moisture, remain to be analysed. A lagged correlation between solar UV exposure and soil water content should exist in dry spells between rainfall events as ET is a flux of water between soil and atmosphere. However, incomplete vegetation cover and low soil water availability frequently constrain ET to slower rates than ET0, probably making the correlations of solar UV exposure with ET and with soil water content weaker than between UV exposure and ET0.

Figure 5 shows the flow of information involved in pre-emptive acclimation to drought. This is a simplification as we have ignored signalling among neighbouring plants—attributed to abscisic acid (ABA) in the soil (Falik et al., 2011)—and the spatial heterogeneity of water availability, which can contribute to pre-emptive acclimation of neighbours of individuals experiencing drought first in a population. This model and the one presented above for the shade avoidance syndrome differ only in the labels, retaining exactly the same structure, which reveals that the generic model in Fig. 3 represents a framework suitable for the study of pre-emptive acclimation under different settings (see effective theory, Box 1).

We can derive three testable hypotheses from this model. (i) If UV exposure triggers pre-acclimation, and this response has evolved as a mechanism for enhancing tolerance of drought, a lagged environmental correlation must exist between solar UV exposure as perceived by plants and future water availability to inform about future drought. (ii) Responses triggered by UV-B+UV-A2 radiation will enhance future tolerance of drought through signalling mechanisms that can be traced to the perception of the cue. (iii) If UV-B and/or UV-A2 radiation function as a purely informational cue, rather than as a stressor, this cue must be perceived through a photoreceptor.

Plants can acquire information on the supply side of their water budget, soil moisture, through their roots, with the hormone ABA being one of the within-plant signals of soil drying (e.g. Tardieu et al., 1992; Wilkinson and Davies, 2010). Further, diffusion of ABA in soil is a signal with potential for plant—plant communication (Novoplansky, 2016), with a putative role in the coordinated regulation of water use among neighbouring
explaining pre-emptive acclimation | 527

plants required for efficient canopy water use (Aphalo, 1991). Taking into consideration that plant roots explore the soil to varying depths, a comprehensive analysis based on the profiles of root length density and soil moisture is needed to assess the relative importance of (i) ABA-mediated sensing of soil moisture, (ii) UV radiation-mediated sensing of evaporative demand, and (iii) integrating soil moisture and evaporative demand as cues for acclimation to future drought. Our model thus leads to the testable prediction that pre-emptive acclimation induced by exposure to solar UV radiation could involve ABA signalling.

Data from an experiment with Arabidopsis, involving exposure to solar UV radiation, but no drought treatment (Rai et al., 2020), can be used to assess if solar UV radiation perceived through the UVR8 photoreceptor affects ABA metabolism and/or signalling. RNA sequencing after 6 h of exposure to different bands of the solar spectrum showed that the abundance of transcripts for several transcription factors responsive to drought or desiccation responded to UV-B and/or UV-A2 radiation in the wild type but not in a mutant lacking functional UVR8. Of these, the transcript abundance of AREB1 (other name ABF2, ABSCISIC ACID RESPONSIVE ELEMENTS-BINDING FACTOR 2) and of GBF3 (G-BOX BINDING FACTOR 3) was increased by exposure to solar UV-B while that of DREB1C (DEHYDRATION RESPONSE ELEMENT-BINDING PROTEIN) was decreased by UV-A2 radiation. For another transcription factor, ATHB7 (ARABIDOPSIS THALIANA HOMEBOX 7), transcript abundance was decreased by exposure to UV-A2, but only in a null mutant lacking the UV-A1+blue light photoreceptors CRY1 and CRY2. ATHB7 is of special interest as it is also responsive to ABA.

Fig. 6. Day to day variation in solar UV radiation and reference evapotranspiration (ET0). Daily sums of estimated ET0 plotted against daily (A) UV-B and (B) UV-A photon exposures. Points indicate daily estimates from observations at 1 min intervals, lines depict the median regression, with grey shading indicating the quartiles (i.e. equivalent to the box in a box plot). Observations are for the growing seasons of years 2020–2021, at Helsinki, Finland (see Aphalo and Sadras, 2021, Preprint).
and has similarity to HaHB4 (*Helianthus annuus* HomeoBox 4), which, as discussed below, when transferred to other crops confers enhanced drought tolerance under field conditions. These responses provide a link between solar UV radiation and the modulation of signalling dependent on ABA and drought.

On the other hand, the abundance of transcripts of DREB1A responded to UV-B radiation both in the wild type and in the UVR8 mutant, suggesting an additional signalling pathway independent of UVR8. However, interestingly, a motif analysis suggests that downstream regulation of expression of genes expected to bind to DREB1A depended on both UVR8 and CRYs. In contrast, neither changes in transcript abundance for genes involved in ABA metabolism nor changes in actual ABA concentration in leaves in response to solar UV radiation could be detected in the same experiment, while transcript abundance for a component of the degradation pathway of ABA, leading to phaseic acid, was responsive.

These results are consistent with the role of UV radiation-induced modulation of ABA signalling influencing readiness to acclimate to drought. Further studies are needed as a role for additional signalling mechanisms can be expected. For a full understanding, sequential measurements through the course of acclimation will be needed. It is also likely that both signalling and end responses differ between phenotypes adapted to different patterns of rainfall and/or evaporative demand (Schwinning and Ehleringer, 2001).

That exposure to solar UV radiation leads to changes in ABA-dependent signalling, a plant hormone which plays a key role in drought tolerance and signalling, supports hypothesis.

Fig. 7. Variation in solar UV radiation and evapotranspiration through the photoperiod. Monthly means for each hour of the photoperiod of estimated reference evapotranspiration plotted against hourly mean (A) UV-B and (B) UV-A photon irradiances. Points indicate monthly estimates from observations at 1 min intervals, lines depict the median regression, with grey shading indicating the quartiles (i.e. equivalent to the box in a box plot). Observations are for the growing seasons of years 2020–2021, at Helsinki, Finland (see Aphalo and Sadras, 2021, Preprint).
(ii) and that most of these changes require functional UVR8, supports hypothesis (iii). We can conclude that a non-stressful, sensory mechanism could enhance drought tolerance in response to solar UV exposure; in other words, an information-driven mechanism conceptually equivalent to anticipatory shade avoidance in response to changes in reflected FR light mediated by phytochromes. This is consistent with the current predominant view that for plants growing in sunlight, exposure to solar UV radiation is rarely a cause of stress (Jansen and Bornman, 2012).

In spite of this evidence for a sensory-driven link between exposure to solar UV radiation and drought tolerance, further experiments are needed to establish the mechanism(s) involved and their ubiquity in both cultivated and wild plants.

Whether further research will fully support or not our hypothesis about the informational role of solar UV radiation in pre-emptive acclimation to drought is not crucial here. The point is that applying our model to this difficult problem allowed us to generate useful and testable hypotheses applicable to both the expected response of plants and the properties of environmental cues. Based on this example, it is possible to imagine how our model will help in assembling the knowledge from different research fields into a broader and deeper understanding of plant phenotypes including pre-emptive acclimation.

Discussion and implications

On how to bridge the gap between laboratory and field

To profit from the mechanistic understanding obtained in controlled environments in natural and farming environments, we need to understand the ecological function of such mechanisms at an equivalent level of detail (Aphalo et al., 2015). At both the mechanistic and ecological levels we need much more than to understand the structure and connections supporting signalling; we need to understand their function also at a higher level of abstraction based on information, taking into consideration both signalling and environmental cues.

If our proposed model holds for multiple cues, one major implication is that metabolic signalling interactions within an organism must reflect the environmental interactions present in the habitats where a species has evolved. Although the rooting volume in potted plants (Poorter et al., 2012) and the spacing between plants growing individually in pots of equal volume and shape (Aphalo and Rikala, 2006) influence growth and morphology, using large pots set at a broad spacing does not solve this problem. Plants grow differently in controlled environments and outdoors (Poorter et al., 2016), and the function of whole-plant canopies depends on responses of individual plants to light and other cues (Maddockni et al., 2002; Pereira et al., 2017). Consequently, full understanding of the role of metabolic signalling unavoidably requires taking into account the ‘normal’ growing environment of each species, even at the level of temporal and spatial variation and correlations among variables. We interpret this as a requirement for molecular and metabolic studies under field conditions, as recently discussed by Schuman and Baldwin (2018), even in the face of the frequently major practical difficulties involved. The gain is, of course, major, as such research will greatly enhance the practical usefulness of a vast amount of data acquired in controlled environments. However, this should not be thought as a competing approach, but as a complementary step, needed for making practical use under field conditions of our ‘how it is implemented’ understanding by developing a detailed understanding of ‘why such signalling or perception mechanism has evolved’ in wild plants and ‘why particular mechanisms have been retained, altered, or lost’ during domestication and breeding in crops. In all cases, quantitative probabilistic multivariate environmental characterizations are essential.

The contribution of pre-emptive acclimation towards plant fitness depends on the dynamics of its regulation. We expect that genetic manipulation to enhance traits such as drought tolerance or yield will most probably succeed through signalling components such as transcription factors or the tuning of sensory systems rather than through direct manipulation of specific physiological traits such as stomatal conductance; for example, the introduction of the gene HaBH4, encoding a transcription factor related to hormonal regulation, has been successful in increasing drought tolerance in crops with only minor trade-offs in the absence of drought (González et al., 2019, 2020). To manipulate traits in this way, we first need to understand how such regulation contributes to yield of crops in the field and to the success of wild plants in specific habitats. This approach can contribute to making science more effective for agriculture, a problem in need of urgent solutions (Passioura, 2020; Sadras et al., 2020).

Ecological and agricultural implications

Plants have evolved sensory mechanisms that allow the acquisition of information from cues and signals, frequently relying on correlations among environmental variables. Climate change is expected to alter the coupling of environmental variables, changing the information they carry. Global warming is altering the relationship between temperature and length of the photoperiod, with implications for both natural and agricultural systems. For example, such altered environmental correlations are important for winter hardening in trees (e.g. Hänninen and Tanino 2011) and crops (e.g. Peltonen-Sainio et al., 2015). Given that different organisms may rely on different cues for timing of developmental events, indirectly, a decoupling among environmental cues may break the synchrony of behaviour and development altering plant–plant, plant–pollinator, and trophic interactions, in this way changing flows of energy and matter in ecosystems (e.g. Brooker, 2006; Salinari et al., 2006; Deacy et al., 2017; Kharouba et al., 2018).

The proposed model provides a unifying theoretical framework for the study of the ecological role of pre-emptive
acclimation in plants, linking environment and plant phenotype across multiple time scales. It has the potential to also contribute to more accurate predictions of the effects of future climate on vegetation.

Conceptual tools to scale molecular understanding to acclimation are also relevant for crop improvement. Current research efforts in plant biology aiming at crop improvement seek to generate more, better, and cheaper genetic and phenotypic data; however, conceptual models of the crop phenotype are lagging (Sadras, 2019). Supporting breeding objectives through the manipulation of the complex signalling pathways involved in metabolic acclimation and regulation processes driven by environmental sensing requires updated models such as the one proposed here. Such regulatory processes can be best understood in relation to the acquisition and use of information. Our model formalizes such an analysis at a high level of abstraction. Less abstract models, specific to pre-emptive acclimation for a given environment and plant species, can be derived from it.

Acknowledgements

Discussions with Rodolfo A. Sánchez, Antonio J. Hall, Frederick Stoddard, Daniel Calderini, Ariel Novoplansky, Carlos L. Ballaré, Jorge J. Casal, T. Matthew Robson, Tarja Lehto, Anders Lindfors, Neha Rai, Yan Yan, Luis O. Morales, and in class with our students have contributed to the ideas presented here. Titta Kotilainen, Anna-Lisa Laine, Tarja Lehto, and Ariel Novoplansky commented on the manuscript. Several students read and commented on an early version of the manuscript as a course assignment, with Susanna Simovaara, Mo Awwanah, Craig Brelsford, Brenda Goh, Duncan Matthews, Marlene Mena Hernández, and Omid Mohammadi making useful suggestions.

Author contributions

PJA: conceptualization with contributions from VOS; PJA and VOS: writing; PJA: visualization.

Conflict of interest

The authors declare that they have no conflicts of interest.

Funding

This work was partly funded by the HiLIFE Grand Challenge programme in Understanding Biological Resilience (BIORESILIENCE), University of Helsinki (grant to PJA).

References

Aasamaa K, Aphalo PJ. 2016. Effect of vegetational shade and its components on stomatal responses to red, blue and green light in two deciduous tree species with different shade tolerance. Environmental and Experimental Botany 121, 94–101.

Aasamaa K, Aphalo PJ. 2017. The acclimation of Tilia cordata stomatal opening in response to light, and stomatal anatomy to vegetational shade and its components. Tree Physiology 37, 209–219.

Ahmad M, Galland P, Ritz T, Wiltschko R, Wiltschko W. 2007. Magnetic intensity affects cryptochrome-dependent responses in Arabidopsis thaliana. Planta 225, 615–624.

Alvarez-Prado S, Sadras VO, Borrás L. 2014. Independent genetic control of maize (Zea mays L.) kernel weight determination and its phenotypic plasticity. Journal of Experimental Botany 65, 4479–4487.

Amir S, Cohen D. 1990. Optimal reproductive efforts and the timing of reproduction of annual plants in randomly varying environments. Journal of Theoretical Biology 147, 17–42.

Anunziata MG, Apelt F, Carillo P, et al. 2017. Getting back to nature: a reality check for experiments in controlled environments. Journal of Experimental Botany 68, 4463–4477.

Aphalo PJ. 1991. Interactions in stomatal function. PhD thesis, University of Edinburgh.

Aphalo PJ. 2010. On how to disentangle the contribution of different organs and processes to the growth of whole plants. Journal of Experimental Botany 61, 626–628.

Aphalo PJ, Ballaré CL. 1995. On the importance of information-acquiring systems in plant–plant interactions. Functional Ecology 9, 5–14.

Aphalo PJ, Ballaré CL, Scopel AL. 1999. Plant–plant signalling, the shade-avoidance response and competition. Journal of Experimental Botany 50, 1629–1634.

Aphalo PJ, Jansen MAK, Mcleod AR, Urban O. 2015. Ultraviolet radiation research: from the field to the laboratory and back. Plant, Cell & Environment 38, 853–855.

Aphalo PJ, Rikala R. 2006. Spacing of silver birch seedlings grown in containers of equal size affects their morphology and its variability. Tree Physiology 26, 1227–1237.

Aphalo PJ, Sadras VO. 2021. Anticipatory responses to drought by plants: what are the environmental cues? EcoEvoRxiv doi: 10.32942/osf.io/yqea. [Preprint].

Badyaev AV. 2005. Role of stress in evolution: from individual adaptability to evolutionary adaptation. In: Hallgrímsson B., Hall BK, eds. Variation: a central topic in biology. Amsterdam: Elsevier, 277–302.

Ballaré CL. 2009. Illuminated behaviour: phytochrome as a key regulator of light foraging and plant anti-herbivore defence. Plant, Cell & Environment 32, 713–725.

Ballaré CL, Pierik R. 2017. The shade-avoidance syndrome: multiple signals and ecological consequences. Plant, Cell & Environment 40, 2530–2543.

Ballaré CL, Sánchez RA, Scopel AL, Casal JJ, Ghera CM. 1987. Early detection of neighbour plants by phytochrome perception of spectral changes in reflected sunlight. Plant, Cell & Environment 10, 551–557.

Baulcombe DC, Dean C. 2014. Epigenetic regulation in plant responses to the environment. Cold Spring Harbor Perspectives in Biology 6, a01947.

Benech Arnold RL, Ghera CM, Sánchez RA, García Fernández AE. 1988. The role of fluctuating temperatures in the germination and establishment of Sorghum halepense (L.) Pers.: regulation of germination under leaf canopies. Functional Ecology 2, 311–318.

Beunza D, Stark D. 2012. From dissonance to resonance: cognitive interdependence in quantitative finance. SSRN Electronic Journal 41, 383–417.

Bloom AJ, Chapin FS, Mooney HA. 1985. Resource limitation in plants: what are the environmental cues? EcoEvoRxiv doi: 10.32942/osf.io/yqea. [Preprint].

Boccalandro HE, Ploschuk EL, Yanovsky MJ, Sánchez RA, Gatz C, Casal JJ. 2003. Increased phytochrome B alleviates density effects on tuber yield of field potato crops. Plant Physiology 133, 1539–1546.

Bourrais G, Burdaki P, Gauthier A, et al. 2015. Large-scale phenomics identifies primary and fine-tuning roles for CRKs in responses related to oxidative stress. PLoS Genetics 11, e1005373.
Explaining pre-emptive acclimation | 5231

Bowsher AW, Shetty P, Anacker BL, Siefert A, Strauss SY, Friesen ML. 2017. Transcriptomic responses to conspecific and congenic competition in co-occurring Trifolium. Journal of Ecology 105, 602–615.

Bradshaw AD. 1965. Evolutionary significance of phenotypic plasticity in plants. Advances in Genetics 13, 115–155.

Brooker RW. 2006. Plant–plant interactions and environmental change. New Phytologist 171, 271–284.

Bruce TJ, Matthes MC, Napier JA, Pickett JA. 2007. Stressful ‘memories’ of plants: evidence and possible mechanisms. Plant Science 173, 603–608.

Casal JJ. 2012. Shade avoidance. The Arabidopsis Book 2012, e0157.

Casal JJ. 2013. Photoreceptor signaling networks in plant responses to shade. Annual Review of Plant Biology 64, 403–427.

Casal JJ, Balasubramanian S. 2019. Thermomorphogenesis. Annual Review of Plant Biology 70, 321–346.

Chan RL, Trucco F, Otegui ME. 2020. Why are second-generation transgenic crops not yet available in the market? Journal of Experimental Botany 71, 6876–6880.

Chapin FS, Matson PA, Vitousek PM. 2011. Principles of terrestrial ecosystem ecology, 2nd edn. Cham: Springer.

Chen TW, Cabrera-Bosquet L, Alvarez Prado S, Perez R, Artzet S, Pradal C, Coupel-Leduc A, Fournier C, Tardieu F. 2019. Genetic and environmental dissection of biomass accumulation in multi-genotype maize canopies. Journal of Experimental Botany 70, 2523–2534.

Chenu K. 2015. Characterising the crop environment: nature, significance and applications. In: Sadras VO, Calderini DF, eds. Crop physiology: applications for genetic improvement and agronomy, 2nd edn. San Diego: Academic Press, 321–348.

Childs DZ, Metcalf CJ, Rees M. 2010. Evolutionary bet-hedging in the real world: empirical evidence and challenges revealed by plants. Proceedings of the Royal Society B: Biological Sciences 277, 3065–3064.

Colwell RK. 1974. Predictability, constancy, and contingency of periodic phenomena. Ecology 55, 1149–1153.

Cossani CM, Sadras VO. 2021. Symmetric response to competition in binary mixtures of cultivars associated with genetic gain in wheat yield. Evolutionary Applications 14, 2064–2078.

Crepy MA, Casal JJ. 2014. Photoreceptor-mediated kin recognition in plants. New Phytologist 205, 329–338.

Dalal A, Attria Z, Moshelion M. 2017. To produce or to survive: how plastic is your crop stress physiology? Frontiers in Plant Science 8, 1010–1024.

Dalal A, Attia Z, Moshelion M. 2017. To produce or to survive: how plastic is your crop stress physiology? Frontiers in Plant Science 8, 1010–1024.

Dawson-Atkins MC, Bergstrom CT, Lachmann M. 2013. When unreliable cues are good enough. The American Naturalist 182, 313–327.

Downes S, Walsh T, Tay WT. 2016. Bt resistance in Australian insect pest species. Current Opinion in Insect Science 15, 78–83.

Dusenbery DB. 1992. Sensory ecology: how organisms acquire and respond to information. New York: W.H. Freeman.

Evans JR. 2013. Improving photosynthesis. Plant Physiology 162, 1780–1793.

Evers JB, Vos J, Andrieu B, Struik PC. 2006. Cessation of tillering in spring wheat in relation to light interception and red/far-red ratio. Annals of Botany 97, 649–658.

Falik O, Hoffmann I, Novoplosky A. 2014. Say it with flowers: flowering acceleration by root communication. Plant Signaling & Behavior 9, e28258.

Falik O, Mordoch Y, Ben-Natan D, Vanunu M, Goldstein O, Novoplosky A. 2012. Plant responsiveness to root–root communication of stress cues. Annals of Botany 110, 271–280.

Falik O, Mordoch Y, Quansah L, Fait A, Novoplosky A. 2011. Rumor has it…: relay communication of stress cues in plants. PLoS One 6, e23625.

Favory JJ, Stec A, Gruber H, et al. 2009. Interaction of COP1 and UVR8 regulates UV-B-induced photomorphogenesis and stress acclimation in Arabidopsis. The EMBO Journal 28, 591–601.

Fisher RA. 1930. The genetic theory of natural selection. London: Oxford University Press.

Fitt G. 1994. Cotton pest management: Part 3. An Australian perspective. Annual Review of Entomology 39, 543–562.

Flack JC. 2017. Coarse-graining as a downward causation mechanism. Philosophical Transactions of the Royal Society A: Mathematical, Physical and Engineering Sciences 375, 20160338.

Furbank RT, Quick WP, Sirault XR. 2015. Improving photosynthesis and yield potential in cereal crops by targeted genetic manipulation: prospects, progress and challenges. Field Crops Research 182, 19–29.

Gagliano M, Renton M, Duvedevani N, Timmins M, Mancuso S. 2012. Out of sight but not out of mind: alternative means of communication in plants. PLoS One 7, e37382.

Geisler M, Gibson DJ, Lindsey KJ, Millar K, Wood AJ. 2012. Upregulation of photosynthesis genes, and downregulation of stress defense genes, is the response of Arabidopsis thaliana shoots to intraspecific competition. Botanical Studies 53, 85–96.

Ghiglione HO, Gonzalez FG, Serrago R, Maldonado SB, Chilcott C, Curé JA, Miralles DJ, Zhu T, Casal JJ. 2008. Autophagy regulated by day length determines the number of fertile florets in wheat. The Plant Journal 55, 1010–1024.

Ghiselin MT. 2000. A bibliography for bioeconomics. Journal of Bioeconomics 2, 233–270.

Gilbert N. 2016. The race to create super-crops. Nature 533, 308–310.

Gitz DC, Liu-Gitz L. 2003. How do UV photomorphogenic responses confer water stress tolerance? Photochemistry and Photobiology 78, 529–534.

Godfrey-Smith P. 2009. Abstractions, idealizations, and evolutionary biology. In: Barberousse A, Morange M, Pradeu T, eds. Mapping the future of biology. Dordrecht: Springer, 47–56.

Godfrey-Smith P. 2013. Philosophy of biology. Princeton, NJ: Princeton University Press.
González FG, Capella M, Ribichich KF, Curin F, Giacomelli JL, Ayala F, Watson G, Otegui ME, Chan RL. 2019. Field-grown transgenic wheat expressing the sunflower gene HAH4B significantly outyields the wild type. Journal of Experimental Botany 70, 1669–1681.

González FG, Rigalli N, Miranda PV, Romagnoli M, Ribichich KF, Trucco F, Portapila M, Otegui ME, Chan RL. 2020. An interdisciplinary approach to study the performance of second-generation genetically modified crops in field trials: a case study with soybean and wheat carrying the sunflower HAH4B transcription factor. Frontiers in Plant Science 11, 178.

Grantham ME, Antonio CJ, O’Neill BR, Zhan YX, Brisson JA. 2016. A case for a joint strategy of diversified bet hedging and plasticity in the pea aphid wing polyphenism. Biology Letters 12, 20160654.

Gruntman M, Novoplansky A. 2004. Physiologically mediated self/ non-self discrimination in roots. Proceedings of the National Academy of Sciences, USA 101, 3863–3867.

Gundel PE, Rudgers JA, Whitney KD. 2017. Vertically transmitted symbionts as mechanisms of transgenerational effects. American Journal of Botany 104, 787–792.

Gutschick VP, BassiriRad H. 2003. Extreme events as shaping physiology, ecology, and evolution of plants: toward a unified definition and evaluation of their consequences. New Phytologist 160, 21–42.

Hänninen H. 2016. Boreal and temperate trees in a changing climate: modelling the ecophysiology of seasonality. Dordrecht: Springer.

Hänninen H, Tanino K. 2011. Tree seasonality in a warming climate. Trends in Plant Science 16, 412–416.

Harper JL. 1977. The population biology of plants. London: Academic Press.

Hayes S, Velanis CN, Jenkins GI, Franklin KA. 2014. UV-B detected by the UVR8 photoreceptor antagonizes auxin signalling and plant shade avoidance. Proceedings of the National Academy of Sciences, USA 111, 11894–11899.

Hebbar KB, Sinha SK. 2002. Effect of external electrical field on mobile water fraction and physiological processes in wheat (Triticum aestivum) leaves. Indian Journal of Experimental Biology 40, 957–959.

Heijde M, Ulm R. 2012. UV-B photoreceptor-mediated signalling in plants. Trends in Plant Science 17, 230–237.

Hides E, Jansen MAK, Strid A. 2013. UV-B exposure, ROS, and stress: inseparable companions or loosely linked associates? Trends in Plant Science 18, 107–115.

Holmes MG, Smith H. 1977a. The function of phytochrome in the natural environment. II. The influence of vegetation canopies on the spectral energy distribution of natural daylight. Photochemistry and Photobiology 25, 539–545.

Holmes MG, Smith H. 1977b. The function of phytochrome in the natural environment. IV. Light quality and plant development. Photochemistry and Photobiology 25, 551–557.

Hyndman R, Athanassopoulos G. 2018. Forecasting: principles and practice, 2nd edn Heathmont, Victoria: O’Texts.

Idânheimo N. 2015. The role of cysteine-rich receptor-like protein kinases in ROS signaling in Arabidopsis thaliana. PhD dissertation, University of Helsinki.

Jackson SD, Thomas B. 1999. The photoperiodic control of plant reproduction. In: Singhal G, Renger G, Sophy S, Irgang Kd, Govindjee, eds. Concepts in photobiology: Dordrecht: Springer Netherlands, 869–896.

Jansen MAK, Borman JF. 2012. UV-B radiation: from generic stressor to specific regulator. Physiologia Plantarum 145, 501–504.

Jarvis PG, McNaughton KG. 1986. Stomatal control of transpiration: scaling up from leaf to region. Advances in Ecological Research 15, 1–49.

Kadam NM, Xiao G, Melgar RJ, Bahuguna RN, Quinones C, Tamiiselvan A, Prasad PV, Jagadish KS. 2014. Agronomic and physiological responses to high temperature, drought, and elevated CO2 interactions in cereals. Advances in Agronomy 127, 111–156.

Karban R. 2015. Plant sensing and communication. Chicago, IL: The University of Chicago Press.

Kauffman SA. 2008. Reinventing the sacred: a new view of science, reason, and religion. Philadelphia: Basic Books.

Kauffman SA. 2016. Humanity in a creative universe. Oxford University Press.

Kharouba HM, Ehrén J, Gelman A, Bolmgren K, Allen JM, Travers SE, Wolkovich EM. 2018. Global shifts in the phenological synchrony of species interactions over recent decades. Proceedings of the National Academy of Sciences, USA 20, 5211–5216.

King EG, Roff DA. 2010. Modeling the evolution of phenotypic plasticity in response to allocation in wing-dimorphic insects. The American Naturalist 175, 702–716.

Krakauer D. 2017. Cryptic genomic nature. In: Walker Sl, Davies PCW, Ellis GRR, eds. From matter to life. Cambridge: Cambridge University Press, 157–173.

Kuijper B, Johnstone RA. 2015. Parental effects and the evolution of phenotypic memory. Journal of Evolutionary Biology 29, 265–276.

Lacaze X, Hayes P, Korol A. 2009. Genetics of phenotypic plasticity: QTL analysis in barley, Hordeum vulgare. Heredity 102, 163-173.

Lyberger KP, Osmond MM, Schreiber SJ. 2021. Is evolution in response to extreme events good for population persistence? The American Naturalist 198, 44–52.

Maddonna GA, Otegui ME, Andrieu B, Chelle M, Casal JJ. 2002. Maize leaves turn away from neighbors. Plant Physiology 130, 1181–1189.

Maffei ME. 2014. Magnetic field effects on plant growth, development, and evolution. Frontiers in Plant Science 5, 445.

Marguerit E, Brendel O, Lebon E, Van Leeuwen C, Ollat N. 2012. Rootstock control of scion transpiration and its acclimation to water deficit are controlled by different genes. New Phytologist 194, 416–429.

Marinho CD, Martins FJO, Amaral Júnior AT, Gonçalves LSA, dos Santos OJAP, Alves DP, Brasileno BP, Peternelli LA. 2014. Genetically modified crops: Brazilian law and overview. Genetics and Molecular Research 13, 5221-5240.

Monteith J, Unsworth M. 2008. Principles of environmental physics, 3rd edn. Academic Press.

Morales LQ, Brousch M, Vainonen J, Jenkins GI, Wargent JJ, Sipari N, Strid A, Lindfors AV, Tegelberg R, Aphalo PJ. 2013. Multiple roles for UV RESISTANCE LOCUS8 in regulating gene expression and metabolite accumulation in Arabidopsis under solar ultraviolet radiation. Plant Physiology 161, 744–759.

Moriconi V, Binkert M, Rojas MCC, Sellaro R, Ulm R, Casal JJ. 2018. Perception of sunflecks by the UV-B photoreceptor UV RESISTANCE LOCUS 8. Plant Physiology 177, 75–81.

Morison J, Baker N, Mullineaux P, Davies W. 2008. Improving water use in crop production. Philosophical Transactions of the Royal Society B: Biological Sciences 363, 5221-5240.

Murphy GP, Acker RV, Rajcan I, Swanton CJ. 2017a. Identity recognition and discrimination of transgenic crops: Brazilian law and overview. Genetics and Molecular Research 13, 2533–2540.

Murphy GP, Swanton CJ, Dudley SA. 2017b. In: Kauffman SA. 2016. Humanity in a creative universe. Oxford University Press.

Murren CJ, Auld JR, Callahan H, et al. 2015. Constraints on the evolution of phenotypic plasticity: limits and costs of phenotype and plasticity. Heredity 115, 293-301.

Noble D. 2012. A theory of biological relativity: no privileged level of causation. Interface Focus 2, 55–64.

Noble D. 2014. Foreword. In: Longo G, Montévil M, eds. Perspectives on or- ganisms: biological time symmetries and singularities. Heidelberg: Springer.

Novoplansky A. 1991. Developmental responses of Portulaca seedlings to conflicting spectral signals. Oecologia 88, 138–140.

Novoplansky A. 2009. Picking battles wisely: plant behaviour under competition. The Plant Cell & Environment 32, 726-741.
Novoplansky A. 2016. Future perception in plants. In: Nadine M, ed. Anticipation across disciplines. Cognitive Systems Monographs. Berlin: Springer.

Novoplansky A, Cohen D, Sachs T. 1990. How Portulaca seedlings avoid their neighbors. Oecologia 82, 490–493.

Orrell D. 2006. The future of everything: the science of prediction. New York: Thunder’s Mouth Press.

Osakabe Y, Yamaguchi-Shinozaki K, Shinozaki K, Tran L. 2013. Sensing the environment: key roles of membrane-localized kinases in plant perception and response to abiotic stress. Journal of Experimental Botany 64, 445–458.

Ovaska J, Walls M, Mutikainen P. 1992. Changes in leaf gas exchange properties of cloned Betula pendula saplings after partial defoliation. Journal of Experimental Botany 43, 1301–1307.

Passiourea JB. 2006. Increasing crop productivity when water is scarce—from breeding to field management. Agricultural Water Management 80, 176–196.

Passiourea JB. 2010. Translational research in agriculture. Can we do it better? Crop and Pasture Science 71, 517–528.

Pedro A, Savin R, Slaper GA. 2012. Crop productivity as related to single-plant traits at key phenological stages in durum field. Field Crops Research 138, 42–51.

Pettonen-Sainio P, Rajala A, Kankaen H, Hakala K. 2015. Improving farming systems in northern Europe. In: Sadras VO, Calderini DF, eds. Crop physiology: applications for genetic improvement and agronomy, 2nd edn. San Diego: Academic Press, 65–91.

Penman HL. 1948. Natural evaporation from open water, bare soil and grass. Proceedings of the Royal Society A: Mathematical and Physical Sciences 193, 120–145.

Pereira ML, Sadras VO, Batista W, Casal JJ, Hall AJ. 2017. Light-mediated self-organization of sunflower stands increases oil yield in the field. Proceedings of the National Academy of Sciences, USA 114, 7975–7980.

Petropoulou Y, Kyparissis A, Nikolopoulos D, Manetas Y. 1995. Enhanced UV-B radiation alleviates the adverse effects of summer drought in two Mediterranean pines under field conditions. Physiologia Plantarum 94, 37–44.

Pierk R, Ballaré CL, Dicke M. 2014. Ecology of plant volatiles: taking a plant community perspective. Plant, Cell & Environment 37, 1845–1853.

Pierk R, de Wit M. 2014. Shade avoidance: phytochrome signalling and other aboveground neighbour detection cues. Journal of Experimental Botany 65, 2815–2824.

Pierk R, Testerink C. 2014. The art of being flexible: how to escape from shade, salt, and drought. Plant Physiology 166, 9–22.

Pigliucci M. 2005. Evolution of phenotypic plasticity: where are we going now? Trends in Ecology & Evolution 20, 481–488.

Poorter H, Bühler J, van Dusschoten D, Climent J, Postma JA. 2012. Size matters: a meta-analysis of the effects of rooting volume on plant growth. Functional Plant Biology 39, 839–850.

Poorter H, Fiorani F, Pieruschka R, Wojciechowski T, van der Putten WH, Kleyer M, Schuur U, Postma J. 2016. Pampered inside, pestered outside? Differences and similarities between plants growing in controlled conditions and in the field. New Phytologist 212, 838–855.

Rai N, Morales LO, Aphiola PJ. 2021. Perception of solar UV radiation by plants: photoreceptors and mechanisms. Plant Physiology 186, 1382–1396.

Rai N, Neugart S, Yan Y, et al. 2019. How do cryptochromes and UVR8 interact in natural and simulated sunlight? Journal of Experimental Botany 70, 4975–4990.

Rai N, O’Hara A, Farkas D, et al. 2020. The photoreceptor UVR8 mediates the perception of both UV-B and UV-A wavelengths up to 350 nm of sunlight with responsiveness moderated by cryptochromes. Plant, Cell & Environment 43, 1513–1527.

Raines CA. 2011. Increasing photosynthetic carbon assimilation in C3 plants to improve crop yield: current and future strategies. Plant Physiology 155, 36–42.

R Core Team. 2021. R: a language and environment for statistical computing. Vienna, Austria: R Foundation for Statistical Computing.

Reymond M, Muller B, Leonard A, Charcosset A, Tardieu F. 2003. Combining quantitative trait loci analysis and an ecophysiological model to analyze the genetic variability of the responses of maize leaf growth to temperature and water deficit. Plant Physiology 131, 664–675.

Reynolds M, Atkin OK, Bennett M, et al. 2021. Addressing research bottlenecks to crop productivity. Trends in Plant Science 26, 607–630.

Reynolds M, Tuberosa R. 2008. Translational research impacting on crop productivity in drought-prone environments. Current Opinion in Plant Biology 11, 171–179.

Robson TM, Hartikainen SM, Aphiola PJ. 2015. How does solar ultraviolet-B radiation improve drought tolerance of silver birch (Betula pendula Roth.) seedlings? Plant, Cell & Environment 38, 953–967.

Ross MS, Flanagan LB, Roi GHL. 1986. Seasonal and successional changes in light quality and quantity in the understory of boreal forest ecosystems. Canadian Journal of Botany 64, 2792–2799.

Ryan CM, Williams M, Grace J, Woolen E, Lehmann CER. 2016. Pre-rain green-up is ubiquitous across southern tropical Africa: implications for temporal niche separation and model representation. New Phytologist 213, 625–633.

Ryan J, Ibiriki H, Sommer R, McNeill A. 2009. Nitrogen in rainfall and irrigated cropping systems in the Mediterranean region. Advances in Agronomy 104, 53–136.

Sadras VO. 2019. Effective phenotyping applications require matching trait platform and more attention to theory. Frontiers in Plant Science 10, 1339.

Sadras VO. 2021. Evolutionary and ecological perspectives on the wheat phenotype. Proceedings of the Royal Society B: Biological Sciences 288, 20211259.

Sadras V, Alston J, Aphiola P, et al. 2020. Making science more effective for agriculture. Advances in Agronomy 163, 153–177.

Sadras VO, Denison RF. 2016. Neither crop genetics nor crop management can be optimised. Field Crops Research 189, 75–83.

Sadras VO, Lake L, Li Y, Farquharson EA, Sutton T. 2016. Phenotypic plasticity and its genetic regulation for yield, nitrogen fixation and δ13C in chickpea crops under varying water regimes. Journal of Experimental Botany 67, 4339–4351.

Sadras VO, Richards RA. 2014. Improvement of crop yield in dry environments: benchmarks, levels of organisation and the role of nitrogen. Journal of Experimental Botany 65, 1981–1995.

Sadras VO, Slaper GA. 2012. Environmental modulation of yield components in cereals: heritabilities reveal a hierarchy of phenotypic plasticities. Field Crops Research 127, 215–224.

Salinari F, Giosue S, Tuliello FN, Rettori A, Rossi V, Spanna F, Rosenzweig C, Gullino ML. 2006. Downy mildew (Plasmopara viticola) epidemics on grapevine under climate change. Global Change Biology 12, 1299–1307.

Schuman MC, Baldwin IT. 2018. Field studies reveal functions of chemical mediators in plant interactions. Chemical Society Reviews 47, 5338–5353.

Schwinnning S, Ehleringer JR. 2001. Water use trade-offs and optimal adaptations to pulse-driven arid ecosystems. Journal of Ecology 89, 464–480.

Sellaro R, Crepy M, Trupkin SA, Karayevok E, Buchovsky AS, Rossi C, Casal J. 2010. Cryptochrome as a sensor of the blue/green ratio of natural radiation in Arabidopsis. Plant Physiology 154, 401–409.

Sellaro R, Pacin M, Casal J. 2012. Diurnal dependence of growth responses to shade in Arabidopsis: role of hormone, clock, and light signaling. Molecular Plant 5, 619–628.

Sellaro R, Yanovscky MJ, Casal J. 2011. Repression of shade-avoidance reactions by sunfleck induction of HY5 expression in Arabidopsis. The Plant Journal 68, 919–928.
Shemesh H, Novoplansky A. 2013. Branching the risks: architectural plasticity and bet-hedging in Mediterranean annuals. Plant Biology 15, 1001–1012.

Shemesh H, Ovadia O, Novoplansky A. 2010. Anticipating future conditions via trajectory sensitivity. Plant Signaling & Behavior 5, 1501.

Shemesh H, Rosen R, Eshel G, Novoplansky A, Ovadia O. 2011. The effect of steepness of temporal resource gradients on spatial root allocation. Plant Signaling & Behavior 6, 1356–1360.

Sinclair TR, Rufty TW, Lewis RS. 2019. Increasing photosynthesis: unlikely solution for world food problem. Trends in Plant Science 24, 1032–1039.

Singh S, Agrawal SB, Agrawal M. 2014. UVR8 mediated plant protective responses under low UV-B radiation leading to photosynthetic acclimation. Journal of Photochemistry and Photobiology B: Biology 137, 67–76.

Smith H. 1981. Light quality as an ecological factor. In: Grace J, Ford ED, Jarvis PG, eds. Plants and their Atmospheric Environment. 21st Symposium of the British Ecological Society. Oxford: Blackwell Scientific Publications, 93–110.

Springer NM. 2013. Epigenetics and crop improvement. Trends in Genetics 29, 241–247.

Stevens M. 2013. Sensory ecology, behaviour, & evolution. Oxford: Oxford University Press.

Stewart BA, Lal R. 2018. Increasing world average yields of cereal crops. Advances in Agronomy 151, 1–44.

Sung S, Amasino RM. 2006. Molecular genetic studies of the memory of winter. Journal of Experimental Botany 57, 3369–3377.

Taiz L, Alkon D, Draguhn A, Murphy A, Blatt M, Hawes C, Thiel G, Robinson DG. 2019. Plants neither possess nor require consciousness. Trends in Plant Science 24, 677–687.

Tardieu F. 2012. Any trait or trait-related allele can confer drought tolerance: just design the right drought scenario. Journal of Experimental Botany 63, 25–31.

Tardieu F, Zhang J, Katerji N, Bettenod O, Palmer S, Davies WJ. 1992. Xylem ABA controls the stomatal conductance of field-grown maize subjected to soil compaction or soil drying. Plant, Cell & Environment 15, 193–197.

Thorogood R, Mustonen V, Aleixo A, et al. 2020. Understanding biologic resilience, from genes to ecosystems. EcoEvoRev doi: 10.32942/osf.io/grpxa. [Preprint].

Tilbrook K, Arongaus AB, Binkert M, Heijde M, Yin R, Ulm R. 2013. The UVR8 UV-B photoreceptor: perception, signaling and response. The Arabidopsis Book 11, e0164.

Trewavas A. 2009. What is plant behaviour? Plant, Cell & Environment 32, 606–616.

Vazquez-Yanes C, Orozco-Segovia A. 1994. Signals for seeds to sense and respond to gaps. In: Caldwell MM, Peary RD, eds. Exploitation of environmental heterogeneity by plants: ecophysiological processes above- and below ground. San Diego: Academic Press, 209–236.

Vigliizzo EF, Frank FC, Carreño LV, Jobbágy EG, Pereyra H, Clatt J, Pincén D, Ricard MF. 2011. Ecological and environmental footprint of 50 years of agricultural expansion in Argentina. Global Change Biology 17, 959–973.

Weiner J. 2019. Looking in the wrong direction for higher-yielding crop genotypes. Trends in Plant Science 24, 927–933.

Weiner J, Du YL, Zhang C, Qin XL, Li FM. 2017. Evolutionary agroecology: individual fitness and population yield in wheat (Triticum aestivum). Ecology 98, 2261–2266.

Weinstein S, Pavlic TP. 2017. Noise and function. In: Walker SI, Davies PCW, Ellis GFR, eds. From matter to life. Cambridge: Cambridge University Press, 174–198.

West-Eberhard MJ. 2003. Developmental plasticity and evolution. Oxford: Oxford University Press.

Wies G, Maddonni G. 2020. Effects of phytochromes B on growth variability and competitive capacity of maize plants in a canopy. Field Crops Research 250, 107765.

Wilkinson S, Davies WJ. 2010. Drought, ozone, ABA and ethylene: new insights from cell to plant to community. Plant, Cell & Environment 33, 510–525.

Wingfield JC. 2008. Organization of vertebrate annual cycles: implications for control mechanisms. Philosophical Transactions of the Royal Society B: Biological Sciences 363, 425–441.

Wolf JB, Wade MJ. 2009. What are maternal effects (and what are they not)? Philosophical Transactions of the Royal Society B: Biological Sciences 364, 1107–1115.

Yan Y. 2021. Long-term exposure to solar blue and UV radiation in legumes: pre-acclimation to drought and accession-dependent responses in two successive generations. PhD thesis, University of Helsinki.

Yan Y, Stoddard FL, Neugart S, Oravec M, Urban O, Sadras VO, Aphalo PJ. 2020. The transgenerational effects of solar short-UV radiation differed in two accessions of Vicia faba L. from contrasting UV environments. Journal of Plant Physiology 248, 153145.

Yan Y, Stoddard FL, Neugart S, Sadras VO, Lindfors A, Morales LO, Aphalo PJ. 2019. Responses of flavonoid profile and associated gene expression to solar blue and UV radiation in two accessions of Vicia faba L. from contrasting UV environments. Photochemical & Photobiological Sciences 18, 434–447.

Yastrebov AB. 1996. Different types of heterogeneity and plant competition in monospecific stands. Oikos 75, 89–97.