



Published in final edited form as:

*Plant Cell Environ.* 2022 April ; 45(4): 1286–1297. doi:10.1111/pce.14281.

## The carbon economics of vegetative phase change

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### Abstract

Across plant species and biomes, a conserved set of leaf traits govern the economic strategy used to assimilate and invest carbon. As plants age, they face new challenges that may require shifts in this leaf economic strategy. In this study, we investigate the role of the developmental transition, vegetative phase change (VPC), in altering carbon economics as plants age. We used overexpression of microRNA 156 (miR156), the master regulator of VPC, to modulate the timing of VPC in *Populus tremula x alba*, *Arabidopsis thaliana* and *Zea mays* to understand the impact of this transition on leaf economic traits, including construction cost, payback time and return on investment. Here, we find that VPC causes a shift from a low-cost, quick return juvenile strategy to a high-cost, high-return adult strategy. The juvenile strategy is advantageous in light-limited conditions, whereas the adult strategy provides greater returns in high light. The transition between these strategies is correlated with the developmental decline in the level of miR156, suggesting that is regulated by the miR156/SPL pathway. Our results provide an ecophysiological explanation for the existence of juvenile and adult leaf types and suggest that natural selection for these alternative economic strategies could be an important factor in plant evolution.

### Keywords

construction cost; leaf economic spectrum; miR156; ontogeny; plant development

## 1 | INTRODUCTION

In the plant world, carbon is queen. It is the currency with which they build, barter and operate. Plants acquire this resource through the enzymatic reactions of photosynthesis which harnesses light energy from the sun to convert CO<sub>2</sub> into sugars. To succeed, plants must photosynthesise efficiently and carbon must be invested wisely.

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### CONFLICT OF INTERESTS

The authors declare that there are no conflict of interests.

### SUPPORTING INFORMATION

Additional supporting information may be found in the online version of the article at the publisher's website.

Leaves are the primary organ through which photosynthesis occurs, and as such, variations in leaf traits that alter carbon economic strategies are of great interest. Leaf construction costs, the amount of carbon required to build leaves, as well as their returns on investment (ROI), determine the resources available for growth and reproduction.

Construction cost and ROI are influenced by the morphological and physiological traits that determine leaf chemical composition, photosynthetic capacity (represented by light saturated photosynthetic rate,  $A_{\text{sat}}$ ), respiration ( $R_d$ ) and leaf lifespan (Poorter et al., 2006; Poorter, 1994). Plants across the globe share conserved relationships between these morphological and physiological traits, creating what is known as the worldwide leaf economics spectrum (LES) (Wright et al., 2004).

The LES uses leaf traits to describe economic strategies ranging from fast and inexpensive to slow and high yielding. Central to the LES is leaf mass per area (LMA), or its inverse, specific leaf area (SLA), which describes ratios between leaf area and mass that quantify changes in leaf thickness and density. Often, the high LMA associated with thicker, denser leaves leads to greater proportions of structural tissue that results in a greater construction cost but also longer leaf lifespan (Poorter et al., 2009; Reich et al., 1992). Because of differences in light interception and proportions of photosynthetic tissues, high-LMA leaves also tend to have lower mass-based  $A_{\text{sat}}$ , leading to a high investment, slow-return economic strategy, with the opposite being true for leaves with low LMA (Reich et al., 1998; Terashima & Hikosaka, 1995; Terashima et al., 2006; Wright et al., 2004). ROI is not directly tied to these strategies, however, as longer lifespan and faster payback of initial cost both have the potential to lead to greater total photosynthetic outputs.

Although largely ignored, ontogenetic variation in leaf economic strategies is equal in magnitude to that between species (Funk et al., 2020; Hayes et al., 2019; Mason et al., 2013). Shifts in leaf traits from those associated with quick-return to longer-return economic strategies are consistently observed with increasing plant age. Further, trait-trait relationships (i.e., the magnitude with which leaf lifespan increases in response to increasing LMA) are significantly altered across plant development, akin to the alterations induced by environment (Damián et al., 2018; Funk et al., 2020; Liu et al., 2019; Niinemets, 2004). These shifts in leaf traits likely have significant ecological impacts altering plant growth, resource acquisition and environmental interactions across its lifespan.

Plant ontogeny includes the juvenile-to-adult vegetative transition known as vegetative phase change (VPC). VPC and its master regulator, microRNA156 (miR156), have been conserved across plant evolution (Axtell & Bowman, 2008). However, the functional significance of this transition and its impacts on fitness, remain major questions in plant developmental biology. As plants progress from the juvenile to adult vegetative phases, the variations in challenges and resources available likely command distinct economic strategies. Previously, we showed that miR156 modulates morphological and physiological traits central to carbon economics (Lawrence et al., 2020). Specifically, the decline in miR156 expression that drives VPC alters SLA, leaf nitrogen (leaf N) and photosynthetic rates across species. miR156-mediated decreases in SLA (equivalent to increases in LMA) between juvenile and adult phases are consistent with the shift from quick to long-return economic strategies

previously described. This suggests miR156 is a regulator of ontogenetic changes in leaf carbon economics and that VPC and the timing of this developmental transition, has important implications for changes in resource use strategies deployed across a plant's lifespan. As a genetically programmed transition, VPC may impact plant fitness by allowing plants to shift between economic strategies as their physiological demands change with age.

In this study, we used wild-type and miR156 overexpressor mutants in three diverse species, *Arabidopsis thaliana*, *Populus tremula x alba* and *Zea mays*, to investigate how VPC—which is driven by a decline in miR156 expression—is related to ontogenetic shifts in carbon economic strategies. We demonstrate that the previously identified phase-specific changes in leaf morphology and photosynthetic physiology lead to shifts from quick to slow-return economic strategies and further show that these strategies are likely to be adaptive under different light environments. The evidence that ontogenetic changes in leaf carbon economics are under the regulation of miR156 not only provides a molecular mechanism for this transition in leaf physiology but also provides an ecophysiological rationale for the existence of vegetative phase change.

## 2 | MATERIALS AND METHODS

### 2.1 | Plant material

*Populus tremula x alba* line 717-1B4 and miR156 overexpressor line 40 described in Lawrence et al. (2021) were obtained by in vitro propagation and hardened on propagation media as described in Meilan and Ma (2006). Plants were then transplanted to Fafard-2 growing mix (Sun Gro Horticulture) in 0.3-L pots in the greenhouse at the University of Pennsylvania (39.9493°N, 75.1995°W, 22.38m a.s.l.) and kept in plastic bags for increased humidity for 2 weeks. Plants were transferred to 4.2-L pots with Fafard-52 growing mix 3 weeks later and fertilised with Osmocote 14-14-14 (The Scotts Company). Plants were additionally fertilised once a week with Peters 20-10-20 (ICL Fertilizers). Greenhouse conditions consisted of a 16-hr photoperiod with temperatures between 22°C and 27°C. Light levels were based on natural light and supplemented with 400-W metal halide lamps (P. L. Light Systems) with daily irradiances between 300 and 1500  $\mu\text{mol m}^{-2} \text{s}^{-1}$  across the day. All settings are controlled by Priva and Microgrow greenhouse systems.

*Z. mays* seeds with the *Corngrass 1 (Cg1)* mutation (stock 310D)—which consists of a tandem duplication of miR156b/c primary sequences described in Chuck et al. (2007)—and the W22 inbred line were obtained from the Maize Genetics Cooperation Stock Center. Plants heterozygous for *Cg1* were crossed to W22 to produce the *Cg1/+* and *+/+* siblings used in this study. Seeds were planted in 9.09-L pots with Fafard-52 growing mix and fertilised with Osmocote 14-14-14 in the greenhouse under growing conditions described above.

*A. thaliana* (Col) and the 35S:miR156 overexpressor line described in Wu and Poethig (2006) were planted in 0.06-L pots with Fafard-2 growing mix. Beneficial nematodes (*Steinernema feltiae*; BioLogic), Marathon® 1% granular insecticide and diatomaceous earth were added to the growing mix to control insects. Planted seeds were placed at 4°C for 3 days before being grown at 22°C in Conviron growth chambers under short days (10 h).

light/14 h. dark) at  $60 \mu\text{mol m}^{-2} \text{s}^{-1}$  light to obtain leaves large enough to fit in the gas exchange chamber. Plants were fertilised with Peters 20-10-20 every other week.

Individuals from genotypes of all species were positioned in a randomised fashion and rotated frequently. Planting was staggered across 2, 3 and 5 months for *A. thaliana*, *P. tremula x alba* and *Z. mays*, respectively.

## 2.2 | Leaf samples

All samples were taken from the uppermost fully expanded leaf. For all three species, naturally juvenile and adult leaves in wild-type lines and 'juvenilised' leaves, those in miR156 overexpressor lines with a juvenile phenotype at leaf positions that would normally be adult, were sampled. In *P. tremula x alba*, developmental stage was determined by petiole shape and abaxial trichome density as described in Lawrence et al. (2021). Juvenile leaves were sampled from wild-type node 10 and adult from node 25 and juvenilised leaves were sampled from overexpressor node 25. In *Z. mays*, developmental stage was determined by the presence or absence of epicuticular wax and trichomes as described in Poethig (1988). Juvenile leaves were sampled from node 4 and adult from node 11 in wild-type plants and juvenilised leaves were sampled from node 11 in *Cg1* mutants. In *A. thaliana*, developmental stage was determined by the presence or absence of abaxial trichomes. Juvenile leaves were sampled from node 5 for physiological and morphological measurements and nodes 2–5 for construction cost measures and adult and juvenilised from node 10 and 10–15 in wild-type and miR156 overexpressors, respectively.

## 2.3 | Leaf construction cost determination

Area of fresh leaf samples was determined from photographs using FIJI software (Schindelin et al., 2012). Samples were then dried at  $60^\circ\text{C}$  until consistent mass, ground using a Willey Mill until small enough to pass through a 2-mm grinding mesh and then ground further using a mortar and pestle. Each *Z. mays* sample consisted of ~100 mg tissue from one leaf, *P. tremula x alba* samples consisted of ~100–120 mg tissue from between one and four leaves and *A. thaliana* samples consisted of ~60–80 mg tissue from around 60 leaves. Chemical composition analysis was performed as described in Cataldo et al. (1975), Poorter (1994) and Poorter and Villar (1997).

One milligram from each sample was used to determine C and N using an ECS 4010 CHNSO Analyzer (Costech Analytical Technologies INC).

For nitrate determination, 20 mg of sample was added to 2 ml  $80^\circ\text{C}$  water for 20 min for nitrate extraction. Samples were centrifuged at 5000 rcf for 15 min. 0.2 ml of supernatant was mixed with 0.8 ml of 5% (w/v) salicylic acid in  $\text{H}_2\text{SO}_4$  and incubated at room temperature for 20 min. Following incubation, 19 ml of 2 N NaOH was added to samples. Absorbance of 410 nm was determined for 0.2 ml aliquots of each sample and  $\text{NO}_3^-$ -N standards of 1 to 200  $\mu\text{g}$ , used to create a standard curve.

The remaining tissue was weighed and used for mineral determination. Samples were ashed at  $550^\circ\text{C}$  in a muffle furnace for 6 h and weighed again. Ash alkalinity was determined in duplicate for each sample to measure  $\text{CO}_3^{2-}$  that formed when oxides from the plant tissue

reacted with CO<sub>2</sub> upon cooling. 4 mg of ash was mixed in 5 ml of deionised H<sub>2</sub>O and two to three drops of 0.5% phenolphthalein were added. The solution was titrated with 0.5 N HCl until the pink indicator color disappeared. An additional volume of HCl, equal to that needed for titration plus an additional 2ml was added to the sample. The solution was then boiled for 5 min, cooled to room temperature and an additional two to three drops of phenolphthalein were added. Samples were then back titrated with 0.5 N NaOH until a faint pink color persisted. The average alkalinity from the two replicates of each sample was used in calculations.

All equations for calculations are presented in the Supporting Information Appendix. Ash alkalinity was calculated using Equation (1.1), mineral content using Equation (1.2) and construction cost of leaf tissue in grams of glucose using Equation (1.3). Tissue mass lost in this process was assumed to have the same chemical composition as that recovered.

## 2.4 | Photosynthetic measurements

All gas exchange measurements were made using a Li-6400 portable photosynthesis machine (Li-Cor Environmental) at a leaf temperature of 25°C following acclimatisation to starting chamber conditions. Light response curves were performed in all three species at a reference [CO<sub>2</sub>] of 400 ppm using a minimum wait time of 2 min between light level changes and data logging. Net photosynthetic rate ( $A_{\text{net}}$ ) in *A. thaliana* was measured at light levels of 1000, 800, 600, 300, 200, 150, 100, 75, 50, 25, 0  $\mu\text{mol m}^{-2} \text{s}^{-1}$  at a flow rate of 300  $\mu\text{mol air s}^{-1}$ , in *Z. mays* at light levels of 1800, 1500, 1200, 1000, 800, 600, 300, 200, 150, 100, 75, 50, 25, 0  $\mu\text{mol m}^{-2} \text{s}^{-1}$  at a flow rate of 400  $\mu\text{mol air s}^{-1}$  and *P. tremula x alba* at light levels of 1500, 1200, 1000, 800, 600, 300, 200, 150, 100, 75, 50, 25, 10 and 0  $\mu\text{mol m}^{-2} \text{s}^{-1}$  at a flow rate of 400  $\mu\text{mol air s}^{-1}$ . Light response curves were analysed using the {AQ Curve fitting} script in R (Tomeo, 2019) which uses equations based on a standard non-rectangular hyperbola model fit described in Lobo et al. (2013) and found in Equation (2.1).

Photosynthetic induction was measured on leaves exposed to light levels less than 20  $\mu\text{mol m}^{-2} \text{s}^{-1}$  for a minimum of 20 min. Induction was measured by logging every 10 s as leaves were exposed to 20  $\mu\text{mol m}^{-2} \text{s}^{-1}$  of light for 2 min and then shifted to a saturating light of 1800  $\mu\text{mol m}^{-2} \text{s}^{-1}$ . Due to the difficulties presented by the small, more delicate nature of *A. thaliana* leaves, induction measurements were not conducted in this species and it is, therefore, not included in any dynamic light modeling.

## 2.5 | Carbon economics calculations

All equations for calculations are described in Poorter (1994) and Poorter et al. (2006) and presented in the Supporting Information Appendix. Assimilation and respiration rates were converted from  $\mu\text{mol CO}_2 \text{ m}^{-2}$  to grams of glucose per gram of tissue using Equation (3.1). SLA used in this equation was calculated by dividing the fresh leaf area by its dry weight. Payback time, the time in days required for leaves to assimilate the equivalent glucose needed to construct it was determined using Equation (4.1). Lifetime return on investment (ROI) was calculated using Equation (4.2) or by modifying this equation using a time span of 7 days for ROI after the first week. Leaf lifespan represents the photosynthetic

lifespan measured as the time between full expansion and senescence, determined by the first signs of discoloration. Payback time and ROI calculations were based on days with 12 h day/night cycles. Assimilation was modeled using Equation (2.1) with the light response curve parameters previously determined. Respiration ( $R_d$ ) used in these equations was estimated as 7%  $A_{\text{sat}}$  to minimise measurement errors that may arise when measuring low gas exchange rates with portable photosynthesis machines as suggested in Poorter et al. (2006). For calculations of payback time and ROI in constant light environments, integrated daily photosynthetic photon flux density (PPFD) was calculated using Equation (5.1). Minimum payback time and maximum lifetime leaf ROI were calculated using the average construction cost and lifespan for each species and developmental phase along with the maximum photosynthetic rate ( $A_{\text{max}}$ ) for each replicate modeled from the light response curves.

## 2.6 | Dynamic light environment model

All parts of the dynamic light environment model were written in R (R Core Team, 2018) and provided as RMarkdown files in the supplement of this manuscript. All equations used are provided in the Supporting Information Appendix.

Part 1 of the model determines light levels across the day based on Campbell and Norman (1998), Zhu et al. (2004) and Salter et al. (2019). Solar declination angle (Equation 6.1), hour angle (Equation 6.2) and solar elevation angle (Equation 6.3) were calculated using a latitude of Philadelphia, PA, USA (39.95°N or 0.697 rads) and Julian day of 180. Direct and diffuse light were calculated using Equations (6.4) and (6.5), respectively. Solar constant was assumed to be 2600  $\mu\text{mol m}^{-2} \text{s}^{-1}$  and atmospheric transmissivity 0.75. Light levels during sun and shade flecks were determined using Equations (6.6) and (6.7), respectively. Leaf area index varied between 0.5 and 8 for each simulation and are reported in Table S5.

Part 2 of the model determines when light switches between sun and shade flecks using Equations (7.1) and (7.2) described in Salter et al. (2019). Day light began at 6:00 and ended at 18:00 with simulations set to begin with a sunfleck. Initial sunfleck lengths varied between simulations and are reported in Table S5.

Part 3 of the model determines assimilation across the day using variables from the light response curves and photosynthetic induction measurements based on Woodrow and Mott (1989), Mott and Woodrow (2000) and Taylor and Long (2017). To determine tau, which describes Rubisco kinetics for photosynthetic induction,  $A_{\text{sat}}$  and instantaneous  $A_{\text{net}}$  during induction were corrected for changes in intercellular  $[\text{CO}_2]$  ( $C_i$ ) using Equations (8.1) and (8.2), respectively.  $C_i$  corrected measures of  $A_{\text{sat}}$  are referred to as  $A_f^*$  and  $A_{\text{net}}$  as  $A^*$ . Tau during increases in light was then calculated as the inverse of the linear slope of  $\ln(A_f^* - A^*)$  versus Time for minutes 1–10 of induction upon exposure to high light (Equation 8.3). The initial minute of induction, often referred to as the ‘fast phase’, was excluded from our model because (1) increases in  $A_{\text{net}}$  during this phase are primarily governed by increases in the pool of RuBP and, therefore, Rubisco kinetics cannot accurately be estimated using gas exchange and (2) at times greater than 1 min, which is the resolution of our model, the contribution of this phase to  $A_{\text{net}}$  is negligible and can be excluded (Woodrow & Mott, 1989). At times greater than 10 min, most changes in  $A_{\text{net}}$  are governed



by stomatal opening and, therefore,  $A^*$  shows little change (Woodrow & Mott, 1989).  $\tau$  during decreases in light describes the deactivation of Rubisco. Because  $A_{\text{net}}$  decreases more quickly than Rubisco deactivation when light levels are reduced,  $\tau$  during deactivation is difficult to estimate using gas exchange. Woodrow and Mott (1989) showed that when measured biochemically,  $\tau$  during deactivation was roughly  $5 \times \tau$  during induction, we, therefore, estimated our values in this way. Induction state, representing the percent of  $A_{\text{sat}}$  instantaneous assimilation is at during induction was calculated by Equation (8.4).

$A_{\text{net}}$  throughout the day required the calculation of the potential maximum assimilation rate ( $A_f$ , Equation 9.1) and initial assimilation rate before induction ( $A_i$ , Equation 9.2) for each 1 min interval as described in Woodrow and Mott (1989) and Taylor and Long (2017).  $\tau$ ,  $A_f$  and  $A_i$  were then used in Equation (9.3), as described in Mott and Woodrow (2000) (Mott & Woodrow, 2000), to calculate  $A_{\text{net}}$  for each time point. Integrated assimilation ( $A_{\text{int}}$ ) across each modeled 1 min interval was calculated by Equation (9.4). To estimate the loss in assimilation due to lags in the response of photosynthesis to light,  $A_{\text{int}}$  with a square response to each change in light was calculated by setting  $\tau$  equal to 0 in Equation (9.4). Subtracting  $A_{\text{int}}$  with induction responses ( $\tau = \text{slope of } \ln(A_f - A^*) \text{ vs Time}$ ) from  $A_{\text{int}}$  with immediate square responses to light ( $\tau = 0$ ) provides the loss in assimilation due to Rubisco activation.

Part 4 of the model uses the same equations described above to calculate payback time and ROI for each simulation. For all modeled calculations, median values for each species and developmental stage were used for assimilation variables and mean values for construction cost and leaf trait variables.

## 2.7 | Statistical analysis

All statistical analyses were performed in JMP<sup>®</sup> Pro v. 14.0.0 (SAS Institute Inc.). Leaf composition, leaf morphology, construction cost and light response curve parameters between adult, juvenile and juvenilised leaves of each species were compared by one-way analysis of variance (ANOVA), where developmental stage was the main effect. When ANOVA results were significant ( $p < 0.05$ ), a Student's  $t$  test was performed to determine differences between developmental groups. Traits were considered to be affected by the developmental phase when adult leaves were significantly different from both juvenile and juvenilised leaves with the same trend. Light induction parameters and leaf lifespan of juvenile and adult leaves of each species were compared using a Student's  $t$  test and considered significantly different when  $p < 0.05$ . Payback time, ROI, photosynthetic rate during induction, induction state and lost assimilation due to slow induction for each species were compared by analysis of covariance with developmental stage as the covariate and considered significantly different when  $p < 0.05$ .

## 3 | RESULTS

### 3.1 | Construction cost is higher for adult than juvenile or juvenilised leaves

The chemical composition of new fully expanded adult, juvenile and juvenilised leaves in three test species was determined to understand how VPC contributes to leaf construction

costs. By using juvenilised leaves in miR156 overexpressor lines (those with a juvenile phenotype at leaf positions that would normally be adult), we are able to separate the effects of VPC from those related to plant size or age. If a measured trait is developmentally phase-specific, juvenilised leaves at 'adult' nodes should be more similar to juvenile leaves than to adult leaves. Per gram of leaf tissue, adult, juvenile and juvenilised leaves require the same amount of glucose ( $p > 0.05$ , Figure S1) in *P. tremula x alba* and *A. thaliana*; thus the composition of leaves is similar across development (Table 1). One exception to this similarity was the concentration of nitrate in *A. thaliana* leaves, which was greater in developmentally juvenile leaves ( $p < 0.05$ ). In *Z. mays*, adult leaves have a greater construction cost per gram of tissue than juvenile and juvenilised leaves due to phase-specific differences in carbon, nitrogen and mineral concentrations ( $p < 0.05$ ) (Figure S1 and Table 1).

At the whole leaf level, adult leaves of all three species cost significantly more ( $p > 0.05$ ) glucose to construct than their juvenile and juvenilised counterparts (Figure 1a–c). This phase-specific pattern is observed even in *P. tremula x alba* and *A. thaliana*, where there are no differences in cost per gram of leaf tissue and is due to the significantly greater area and mass of adult leaves compared to both juvenile and juvenilised leaves ( $p > 0.05$ ) (Table S1). Overall, we find that regardless of differences in chemical composition, the effect of VPC on leaf size leads to higher costs for adults compared to juvenile leaves.

### 3.2 | Leaf payback time becomes longer as plants transition from juvenile to adult, but the difference in magnitude depends on light environment

Payback time, the amount of time it takes for a leaf to assimilate the carbon originally invested in its construction is greater for adult leaves than juvenile or juvenilised leaves across all light levels for all three test species (Figure 1d–f). Interestingly, there is a significant interaction between developmental phase and light ( $p < 0.05$ ) in all three species as the difference between developmental phases is greater as light decreases. Specifically, adult leaves of *P. tremula x alba* have, respectively, 7.85- and 3.4- fold longer payback time than juvenile and juvenilised leaves under low light ( $10 \mu\text{mol m}^{-2} \text{s}^{-1}$ ), but only 6.6- and 1.79-fold longer payback time under high light ( $1000 \mu\text{mol m}^{-2} \text{s}^{-1}$ ). In *A. thaliana*, the payback time for adult leaves is, respectively, 18.8- and 5.7-fold longer under low light and 9.4 and 3.16-fold longer under high light, than for juvenile and juvenilised leaves. Lastly, in *Z. mays*, adult leaves have, respectively, 8.13 and 11.64-fold longer payback time under low light and 6.3 and 12-fold longer payback time under high light than juvenile and juvenilised leaves. Because construction cost remains constant across light levels in our modeled payback time, these differences are a result of photosynthetic responses to light modeled using light response curves (Table S2). Similar relationships between payback time and light are observed on a per gram basis for all three species, although to a lesser extent (Figure S1). As we saw with construction costs, the greater similarity between juvenilised leaves and juvenile leaves, as opposed to adult leaves, indicates differences in miR156-mediated development, rather than plant size or age, is responsible for the payback differences observed here.



### 3.3 | Light environment alters the phase-specific relationship of leaf return on investment

Despite higher construction costs and longer payback time, return on investment (ROI) after 1 week is higher for adult leaves than for juvenile and juvenilised leaves in high-light environments (Figure 1g–i). Adult leaves outperform juvenile leaves in high-light environments (i.e., PAR > 340, 400 and 90  $\mu\text{mol m}^2 \text{s}^{-1}$  for *P. tremula x alba*, *A. thaliana* and *Z. mays*, respectively). In low-light environments, however, adult leaves produce fewer returns than juvenile leaves and can even experience net carbon loss (i.e., PAR < 270, 100 and 80  $\mu\text{mol m}^2 \text{s}^{-1}$  for *P. tremula x alba*, *A. thaliana* and *Z. mays*, respectively). Furthermore, the ROI for juvenile and juvenilised leaves is less sensitive to light environment than is the case for adult leaves. Juvenile and juvenilised leaves of *P. tremula x alba* and *Z. mays* approach their maximum ROI at a PPFD around 10  $\text{mol m}^{-2} \text{day}^{-1}$ . For *A. thaliana*, which was grown under low-light conditions for this experiment, the ROI for adult leaves continues to increase as PPFD increases, well past 40  $\text{mol m}^{-2} \text{day}^{-1}$ . ROI per gram of tissue display similar patterns to leaf-based measures in *P. tremula x alba* and *Z. mays* across light environments. However, in *A. thaliana*, juvenile and juvenilised tissue maintain a higher ROI than adult tissue across all light environments (Figure S1).

### 3.4 | Leaf lifespan is longer for adult leaves than juvenile leaves

To understand how phase-specific differences in leaf construction cost and payback time impact the overall economic strategy of juvenile and adult leaves, we also measured the photosynthetic lifespan of these leaves. Previously, we determined that SLA, which is closely connected to lifespan, is a phase-specific trait, with adult leaves having lower SLA (or higher LMA) compared to both juvenile and juvenilised leaves of all three test species (Table S1) (Lawrence et al., 2020). Across species, thicker, more dense adult leaves (low SLA) had significantly longer ( $p < 0.05$ ) lifespans than juvenile leaves (Figure 2a–c). Lifespan differences between juvenile and adult leaves ranged from 26 to 23 days in *P. tremula x alba* and *Z. mays*, respectively, to 8 days in *A. thaliana*. Of note, the low-light payback time for adult *Z. mays* leaves (which approaches 52 days) far exceeds the 34-day average lifespan of these leaves. The longer lifespan of adult leaves results in a significantly greater ( $p < 0.05$ ) maximum lifetime ROI for adult leaves compared to juvenile leaves of all three species (Figure 2d–f).

### 3.5 | Photosynthetic induction is faster in juvenile leaves than in adult leaves

Because light levels continuously fluctuate throughout the day in forests and crop fields (i.e., from the sun moving across the sky, leaves fluttering in the wind, etc.), modeling carbon economics of *P. tremula x alba* and *Z. mays* under more realistic conditions required an analysis of the rate of photosynthetic induction in juvenile and adult leaves. Upon exposure to saturating light, juvenile leaves of both species more quickly reached higher photosynthetic rates and induction states than adult leaves (Figure 3 and S3). The relaxation times for Rubisco activation were not significantly different ( $p < 0.05$ ) between juvenile and adult leaves (Table S3), indicating that developmental differences in induction occur before 1 min, during the ‘fast-phase’ of induction.

Alternatively, developmental differences in induction could be due to a combination of traits, such as stomatal conductance and Calvin cycle intermediate accumulation, rather than solely

to the activation of Rubisco. In *P. tremula x alba*, differences in photosynthetic induction are apparent by 1 min of high-light exposure, suggesting phase-specific differences in induction are likely to present before exposure to high light, or during the first minute when buildup of intermediates in the Calvin cycle is most important. This is not the case for *Z. mays*, where juvenile leaves reach a higher induction state than adult leaves during this ‘slow-phase’ period of minutes 1–10 (Figure S3). These results suggest that while juvenile leaves of both species have faster photosynthetic induction properties than their adult counterparts, the mechanisms behind these differences may vary between species.

### 3.6 | Dynamic light models show phase-specific leaf economic relationships are more dependent on daily light than number of sunflecks

For both *P. tremula x alba* and *Z. mays*, the dynamics of light environment had a significant effect ( $p < 0$  on the relationship of payback time and ROI between juvenile and adult leaves (Figure 4). Over 156 different light simulations, daily PPFD varied between 18 and 32 mol m<sup>-2</sup> day<sup>-1</sup> while the number of sunflecks varied between 1 and 242 (figures of simulated light environments in Figure S2). The carbon economic traits of adult leaves were much more affected by PPFD compared to juvenile leaves, as indicated by the smaller slope of the negative relationship with payback time and the larger slope of the positive relationship with ROI for adult leaves of both species (Table S4). Similar trends are observed on a per gram of tissue basis, although differences between juvenile and adult leaf payback time across PPFD are not significant ( $p > 0.05$ ) for *P. tremula x alba* (Figures S4 and S5 and Table S4).

Surprisingly, there was no significant interaction between developmental phase and the number of sunflecks for payback time or ROI in *P. tremula x alba* (Figure S3 and Table S4). While significant interactions ( $p < 0.05$ ) were present for these relationships in *Z. mays*, the low  $R^2$  values for both developmental phases and economic traits ( $R^2 = 0.05$ ), indicate that sunflecks have a minor effect on payback time and ROI (Table S4). There were no significant differences in the way juvenile and adult tissue responded to sunflecks on a per gram basis and any significant relationships between payback time or ROI and sunflecks for either developmental stage were minor ( $R^2 = 0.05$ ) (Figures S4 and S5 and Table S4).

Despite there being no meaningful relationship between carbon economics and the number of sunflecks, lags in photosynthetic response to light fluctuations due to the rate of induction resulted in assimilation loss for both developmental phases in both species. As the number of sunflecks increased and plants were exposed to more rapid changes in light, the assimilation lost due to a lag in induction also increased (Figure 3). In both species, the faster induction rate in juvenile leaves resulted in lower assimilation losses compared to adult leaves ( $p < 0.05$ ). Nevertheless, the impact of these losses on carbon economics in these simulated environments is minimal compared to the effect of overall changes in PPFD.

## 4 | DISCUSSION

Vegetative phase change alters plant economic strategies through miR156-mediated changes in leaf morphology and physiology. Juvenile leaves—which have high levels of miR156—use a low-cost, quick-return economic strategy, whereas adult leaves—which have low levels of miR156—use a high-cost, slow-return strategy. The adult strategy carries more

risk than the juvenile strategy but has the potential to provide high ROI (Figure 5). This developmental shift in strategy is brought about by the same traits that govern leaf economics across species and environments in the LES, namely, leaf lifespan and LMA (Wright et al., 2004). Across both C<sub>3</sub> (*A. thaliana* and *P. tremula x alba*) and C<sub>4</sub> (*Z. mays*) species, adult leaves have high LMA and long lifespan while juvenile leaves have low LMA and shorter lifespan (Figure 2a–c and Table S1). In *Z. mays*, leaf N and the photosynthetic rates of juvenile and adult leaves follow the established trait relationships within the LES, as the low LMA juvenile leaves also have higher mass-based measures of N and  $A_{\text{sat}}$  compared to adult leaves (Table 1) (Lawrence et al., 2020). As previously reported, trait-trait relationships of the LES are not always conserved at smaller than global scales (Anderegg et al., 2018; Edwards et al., 2014; Mason & Donovan, 2015b). We find this to be the case for developmental changes in leaf N and  $A_{\text{sat}}$  in *P. tremula x alba* and *A. thaliana*, as these leaves have no significant differences in mass-based measures of N and  $A_{\text{sat}}$  despite their differences in LMA. It is unclear why the expected negative relationships between LMA and leaf N or  $A_{\text{sat}}$  are lacking, as LMA increases during VPC in these species, although it is not likely to be related to differences in photosynthetic pathways as both C<sub>3</sub> and C<sub>4</sub> species are included in global datasets. However, previous work found no phase-specific changes in photosynthetic nitrogen use efficiency (PNUE) (Lawrence et al., 2020), indicating adult leaves somehow compensate for the structural changes that often reduce PNUE in high-LMA leaves, potentially through their increased stomatal density which could reduce resistance to CO<sub>2</sub> diffusion (Feng et al., 2016; Hikosaka, 2004; Lawrence et al., 2021).

These developmentally programmed changes in leaf carbon economic strategy are likely to have ecological implications because plants face different biotic and abiotic challenges during their lifetime. For example, juvenile leaves, which have a photosynthetic advantage over adult leaves under low light (Lawrence et al., 2020), are more likely to be found in low, highly dynamic, light environments because young plants are often shaded by their neighbors and quickly self-shade due to their relatively rapid rate of leaf production (Wang et al., 2008). Here, we find that the economic strategy of juvenile leaves further adds to this low-light advantage as these leaves are able to maintain a positive carbon balance even at very low light levels (Figures 1 and 5). The payback time of adult leaves dramatically increases under low light, greatly reducing ROI and, in some cases, exceeds the lifespan of a leaf, resulting in net carbon loss. On the contrary, the magnitude with which adult ROI exceeds that of juvenile leaves increases significantly with increasing irradiance (Figure 1). Overall, the economic strategy of juvenile leaves appears to be less sensitive to light environment, making it a low-risk, low-reward strategy that is likely beneficial for a young plant with minimal carbon reserves. Although the high-cost strategy of adult leaves incurs greater risk because of long-term environmental variability, the high-reward potential of this strategy may outweigh this risk.

Surprisingly, the ability of juvenile leaves to respond more quickly to sunflecks than adult leaves had little effect on the carbon economic relationships between these leaves in our dynamic light models (Figure S3). In our simulated environments, we held the total time leaves were exposed to sunflecks relatively constant but allowed the number of sunflecks to vary dramatically. As a result, there was no correlation between the number of sunflecks

and daily integrated PPFD. It may be that developmental differences in induction rate have a greater influence on carbon economic relationships when sunflecks play a large role in determining daily PPFD, such as in a rain forest understory where sunflecks can account for 52% of daily light (Chazdon & Pearcy, 1991).

Notably, the developmental shift in leaf carbon economics caused by VPC is consistent across species with both C<sub>3</sub> and C<sub>4</sub> photosynthetic pathways. The C<sub>4</sub> carbon concentrating mechanism requires added energy from light reactions but reduces the requirements for photosynthetic proteins and carbon lost to photorespiration, often leading to faster growth over C<sub>3</sub> species (Ehleringer & Cerling, 2002; Long, 1999). This allows C<sub>4</sub> plants with the same leaf functional traits (i.e., LMA) as a C<sub>3</sub> species to assimilate more carbon. However, there is no interaction between photosynthetic pathway and leaf economic traits (Simpson et al., 2020), meaning that a shift from fast to slow economic traits, as we find between juvenile and adult leaves, will affect the carbon economics of C<sub>3</sub> and C<sub>4</sub> species to the same extent, consistent with our findings here.

The developmental differences in carbon economics described here indicate that genotypic variation in miR156 expression and subsequently the timing of VPC could have significant consequences for plant ecology and evolution. Among other things, leaf economic strategies alter plant growth and survival in response to nutrient and water availability, herbivory, competition and light environment (Chen et al., 2020; Coley, 1988; Mason & Donovan, 2015a; Poorter et al., 2006; Reich, 2014; Russo & Kitajima, 2016). That the developmentally regulated changes in leaf economic strategy are conserved among juvenile and adult leaves of three phylogenetically diverse species and that this strategy confers the ability to respond to changes in environmental factors, suggests these ontogenetic changes in carbon economics are widely advantageous. A better understanding of natural variation in the timing of VPC and the function of this process in plant physiology and response to environmental stressors is crucial for determining the role of this developmental transition in plant ecology and evolution.

## Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

## ACKNOWLEDGEMENTS

We thank Samara Gray and Joshua Darfler for their assistance in caring for the plants used in this study. This study was funded by the NSF Graduate Research Fellowship (Division of Graduate Education; DGE-1845298), University of Pennsylvania SAS Dissertation Research Fellowship and the Peachey Research Fund awarded to Erica H. Lawrence and NIH GM51893 awarded to R. Scott Poethig.

## Funding information

National Science Foundation, Grant/Award Number: DGE-1845298; National Institutes of Health, Grant/Award Number: GM51893

## DATA AVAILABILITY STATEMENT

The data that support the findings of this study are available from the corresponding author upon reasonable request.

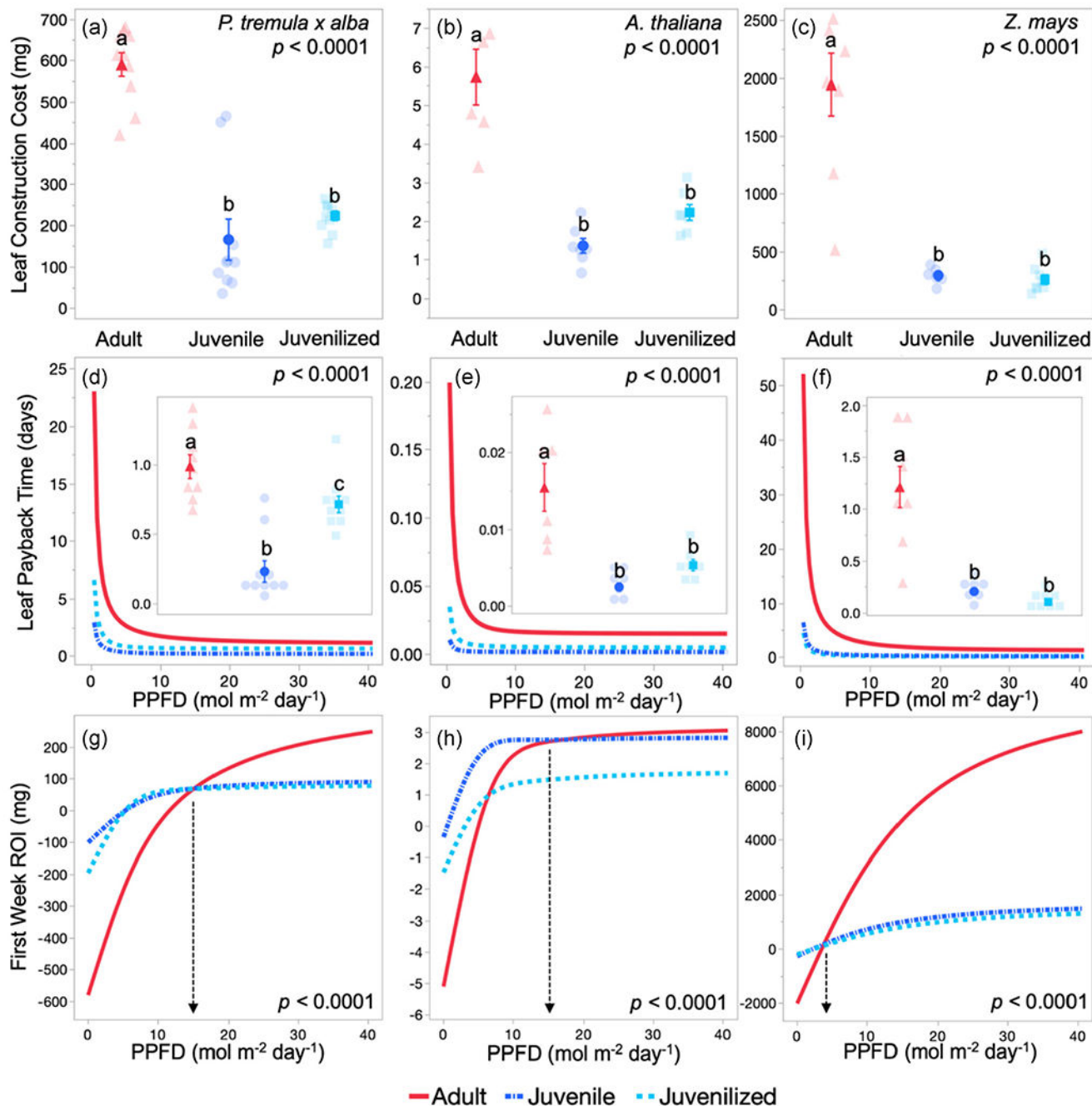
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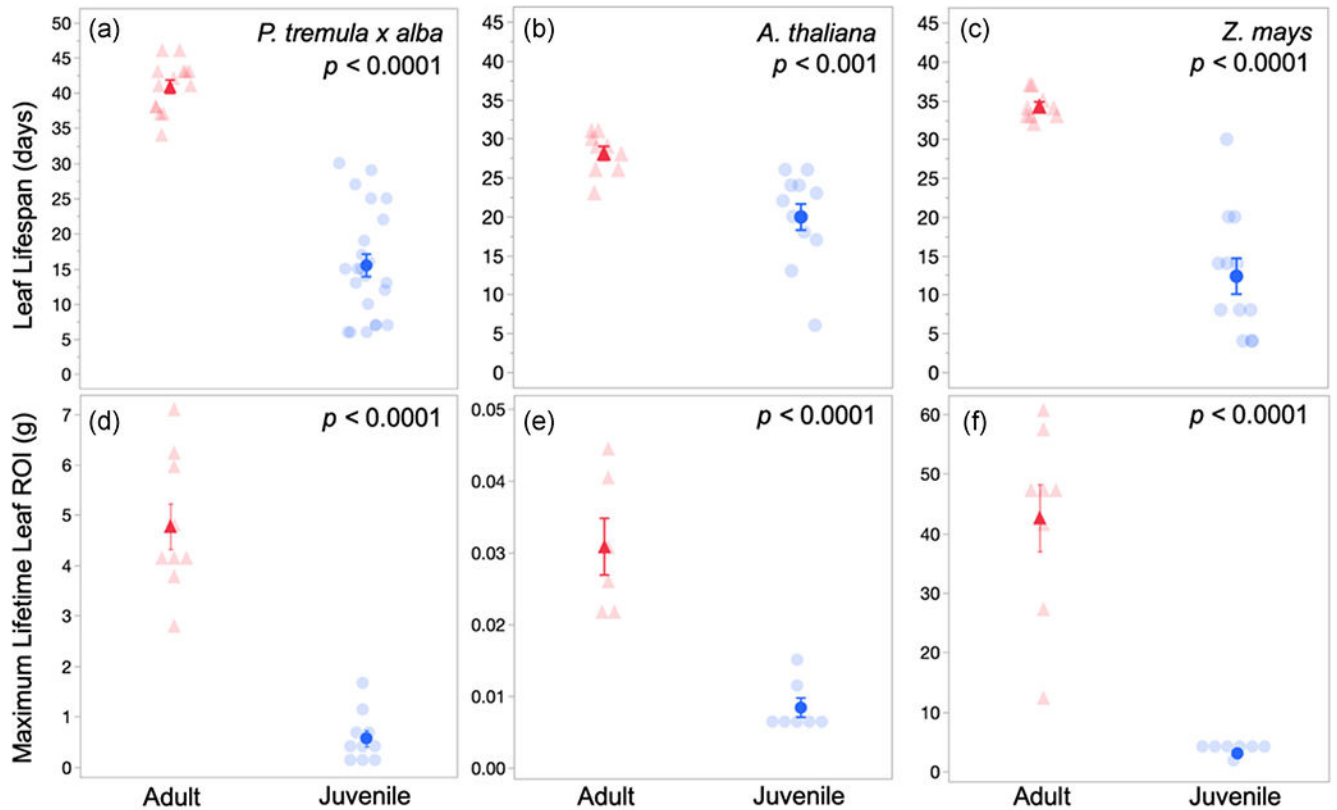
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**FIGURE 1.**

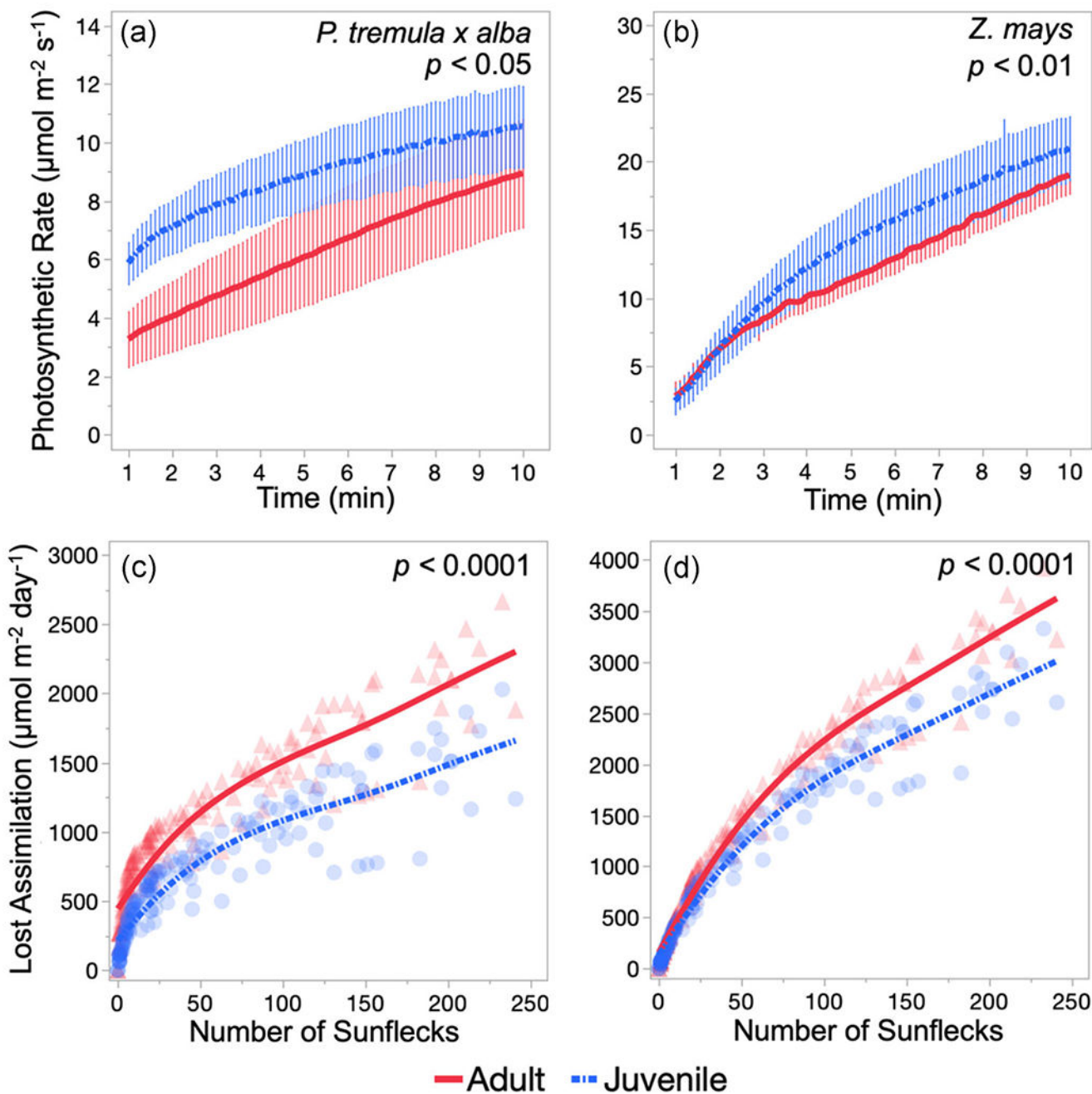
Leaf construction cost in grams of glucose (a–c), Leaf payback time in days (d–f) and return on investment (ROI) after the first week for adult (red triangles and solid lines), juvenile (blue circles and dash-dotted lines) and juvenilised (light blue squares and dashed lines) leaves of *Populus tremula x alba* (a,d,g), *Arabidopsis thaliana* (b,e,h) and *Zea mays* (c,f,i). Insets in (d–f) depict minimum leaf payback time calculated using  $A_{max}$  modeled from light response curves. Construction cost and minimum payback time are presented as means  $\pm$  SEM by solid symbols and individual replicates by transparent symbols.  $p$  Values

determined by one-way analysis of variance with leaf development as the effect. Different lower-case letters represent groups significantly different from each other as determined by Student's *t* test. Payback time and ROI are modeled using photosynthetic light response parameters and 12-h light periods consisting of constant PAR levels between 10 and 940  $\mu\text{mol m}^{-2} \text{s}^{-1}$  and are plotted against the resulting daily integrated photosynthetic photon flux density (PPFD) (i.e., a leaf in a low light environment with PAR 10 would experience  $\sim 0.47 \text{ mol m}^{-2} \text{ light day}^{-1}$ ). *p* Values determined by analysis of covariance (d–i). Vertical arrows in (g–i) indicate the light level where adult leaves begin to have a higher ROI than juvenile leaves during the first week

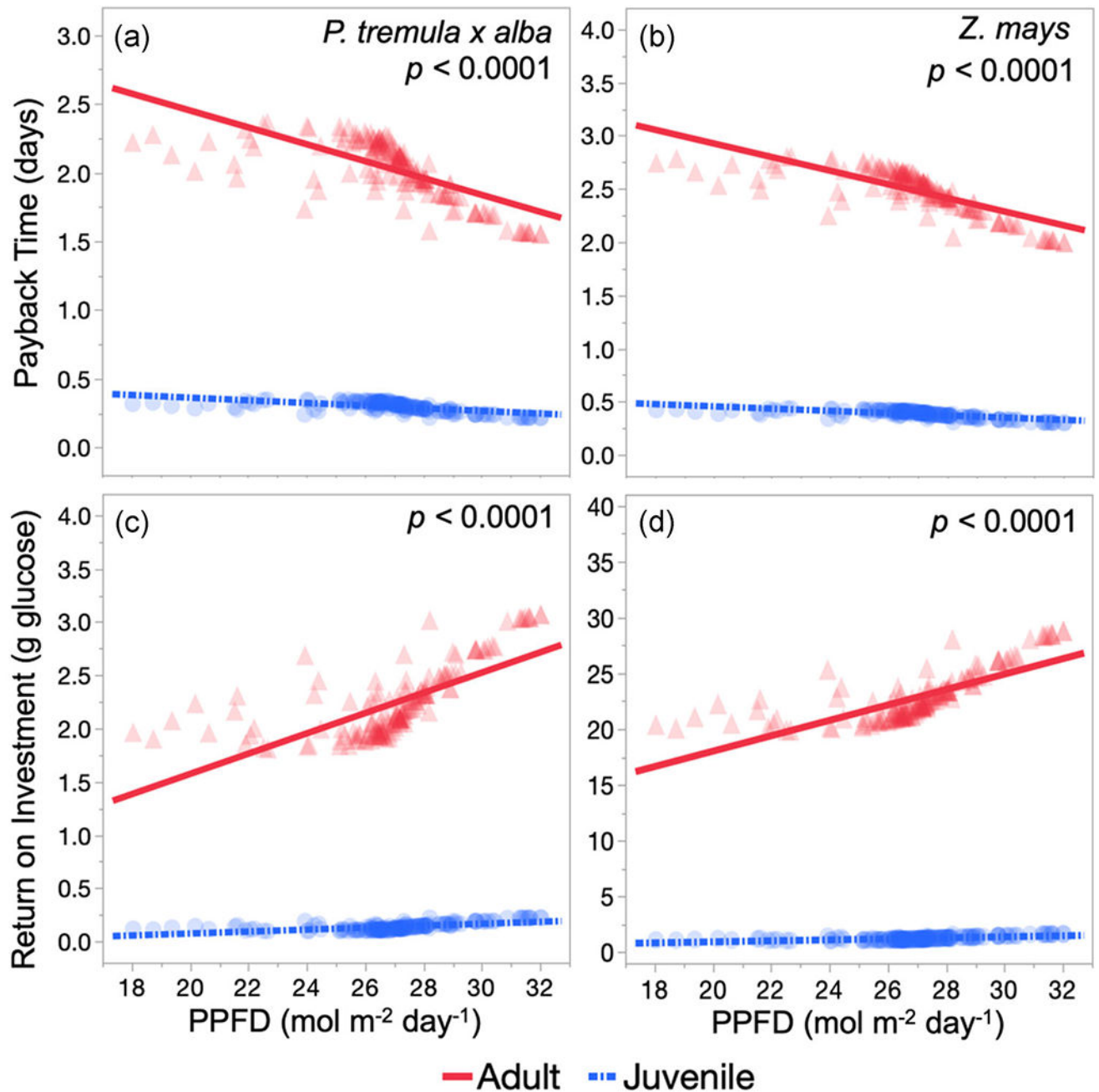


**FIGURE 2.**

Leaf lifespan (a–c) and maximum lifetime return on investment (ROI) per leaf in grams of glucose (d–f) for adult (red triangles) and juvenile (blue circles) leaves of *Populus tremula x alba* (a,d), *Arabidopsis thaliana* (b,e) and *Zea mays* (c,f). Maximum lifetime ROI calculated using average leaf lifespan for each species and developmental phase and  $A_{\max}$  modeled from light response curves. Data presented as means  $\pm$  SEM by solid symbols and individual replicates by transparent symbols.  $p$  Values determined by Student's  $t$  test

**FIGURE 3.**

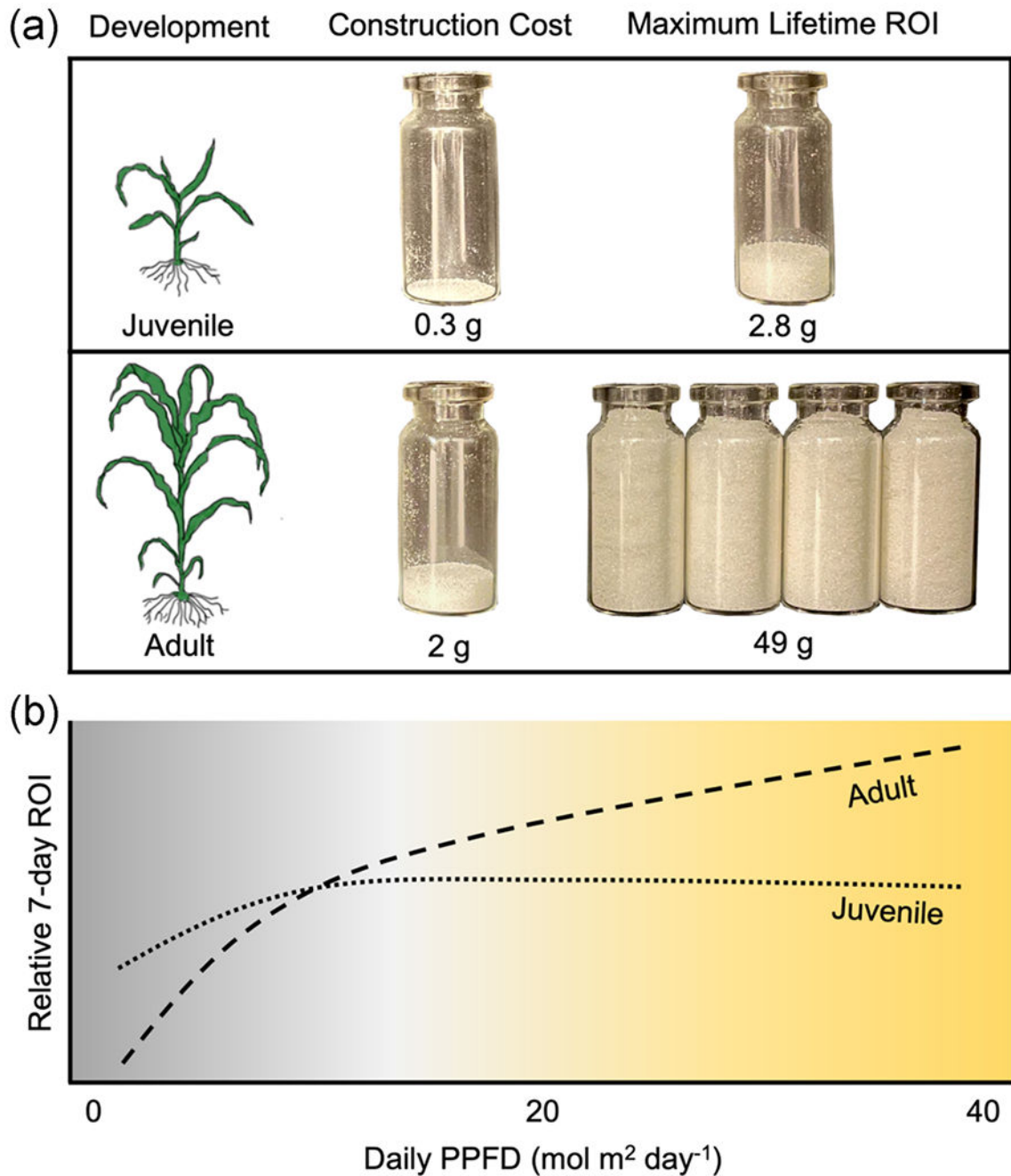
Photosynthetic rate during minutes 1–10 of light induction upon exposure to saturating light (a,b) and the assimilation lost due to slow photosynthetic induction in modeled light environments that vary in number of sunflecks across the day (c,d) for adult (solid, red) and juvenile (dash-dotted, blue) leaves of *Populus tremula x alba* (a,c) and *Zea mays* (b,d). Photosynthetic rate presented as the mean  $\pm$  SEM and lost assimilation presented as individual replicates by transparent symbols and smoothed mean line.  $p$  Values determined by analysis of covariance



**FIGURE 4.**

Leaf payback time (a,b) and lifetime return on investment (c,d) for adult (red triangles and solid lines) and juvenile (blue circles and dash-dotted lines) leaves from simulated dynamic light environments in *Populus tremula x alba* (a,c) and *Zea mays* (b,d) plotted against daily integrated photosynthetic photon flux density (PPFD). Data presented as individual replicates by transparent symbols and linear line of best-fit. *p* Values determined by analysis of covariance



**FIGURE 5.**

Juvenile leaves have a low-cost, low-return carbon economic strategy best suited for low-light environments while adult leaves have a high-cost, high-return strategy better suited for high-light environments. Visual representations of the differences in construction cost and maximum lifetime return on investment (ROI), in grams of sugar, between juvenile and adult leaves of *Zea mays* (a). Maximum lifetime ROI was calculated using leaves photosynthesizing at light-saturated photosynthetic rates for 12 h a day during their full lifespan. Panel (b) shows the relative 7-day ROI for juvenile and adult leaves of *Z. mays*

across light environments, depicting the advantage for the juvenile strategy in low light ( $<10 \text{ mol m}^2 \text{ day}^{-1}$ ) and adult leaves in high-light environments

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Leaf composition of adult, juvenile and juvenilised leaves of *Populus tremula x alba*, *Arabidopsis thaliana* and *Zea mays*

TABLE 1

Species	Development	Carbon (mg g <sup>-1</sup> )	Nitrogen (mg g <sup>-1</sup> )	Nitrate (mg g <sup>-1</sup> )	Mineral (mg g <sup>-1</sup> )	n
<i>P. tremula x alba</i>	Adult	793.96 ± 29.96	37.06 ± 2.24	0.14 ± 0.03 <sup>b</sup>	168.97 ± 30.86 <sup>b</sup>	5
	Juvenile	730.22 ± 69.20	63.22 ± 11.25	0.31 ± 0.10 <sup>a</sup>	206.56 ± 59.56 <sup>b</sup>	4
	Juvenilised	618.49 ± 72.95	72.92 ± 17.19	0.21 ± 0.04 <sup>b</sup>	308.58 ± 59.60 <sup>a</sup>	5
<i>A. thaliana</i>	Adult	547.87 ± 72.32	106.03 ± 28.01	2.10 ± 0.39 <sup>b</sup>	346.1 ± 69.95	3
	Juvenile	564.59 ± 10.70	98.11 ± 27.55	2.58 ± 0.24 <sup>a</sup>	337.29 ± 37.08	3
	Juvenilised	536.19 ± 22.86	123.82 ± 7.97	2.46 ± 0.21 <sup>a</sup>	339.98 ± 30.77	3
<i>Z. mays</i>	Adult	772.74 ± 5.21 <sup>a</sup>	24.73 ± 3.43 <sup>b</sup>	0.09 ± 0.01 <sup>b</sup>	202.52 ± 3.26 <sup>c</sup>	5
	Juvenile	620.09 ± 5.69 <sup>b</sup>	42.86 ± 4.82 <sup>a</sup>	0.30 ± 0.04 <sup>a</sup>	337.05 ± 8.44 <sup>a</sup>	5
	Juvenilised	687.2 ± 14.14 <sup>c</sup>	43.77 ± 1.33 <sup>a</sup>	0.17 ± 0.02 <sup>b</sup>	269.03 ± 14.00 <sup>b</sup>	5

Note: Values indicate the mean ± standard error. For traits where results of the analysis of variance were significant ( $p < 0.05$ ), lower case letters indicate significant ( $p < 0.05$ ) differences between developmental groups determined by a Student's *t* test.