Combining modelling and experimental approaches to assess the feasibility of developing rice-oil palm agroforestry system

Raphaël P.A Perez[1], Rémi Vezy[2], Romain Bordon[1], Thomas Laisné[1], Sandrine Roques[1], Maria-Camila Rebolledo[1], Lauriane Rouan[1], Denis Fabre[1], Olivier Gibert[1], Marcel De Raissac[1]

[1]CIRAD, UMR AGAP, F-34398 Montpellier, France.

UMR AGAP Institut, Univ Montpellier, CIRAD, INRAE, Institut Agro, F-34398 Montpellier, France.

[2]CIRAD, UMR AMAP, F-34398 Montpellier, France.

AMAP, Univ Montpellier, CIRAD, CNRS, INRAE, IRD, Montpellier, France

For correspondence: raphael.perez@cirad.fr

Keywords: agroforestry, light intensity, light fluctuation, rice, shading, yield

Abstract

Climatic hazards affecting the main rice producing regions of Indonesia increase the risk of annual production loss and encourage the development of innovative strategies to maintain stable production. Conversion of oil palm monocultures to rice-based intercropping systems is a strategy to be considered, but relies on the existence of suitable planting management that optimizes both palm productivity while providing enough light for undergrowth rice varieties tolerant to shady conditions. This paper proposes to couple a model of light interception on virtual canopies with indoor experiments to evaluate the feasibility of developing rice-oil palm agroforestry systems. We first selected a planting design that optimized the transmitted light available for rice using a functional-structural plant model (FSPM) of oil palm. Secondly, we reproduced the light regime simulated with specific changes in the intensity and the daily fluctuation of light in controlled conditions. Three light treatments were designed to test independently the effect of daily light quantity and the effect of diurnal fluctuation on contrasted rice subpopulations.

Light quantity was the main factor driving changes in plant morphology and architecture, while light fluctuation only appeared to explain variations in yield components and phenology. This study highlighted the importance of light fluctuation in the grain filling process and resource reallocation.

The conservation of relative change among varieties between treatments suggests that varietal responses to low light are likely to be heritable, and that varietal screening under full light can provide clue on varietal behavior under low light. However, the identification of specific traits such as a limited expansion of leaf area and a conservation of leaf senescence under shade and high light fluctuation paves the way for selecting varieties dedicated to agroforestry systems. Further investigations including light quality and larger genotypic population to screen are discussed.

Introduction

The recurrent droughts caused by the 'El Niño' phenomenon have a significant impact on both rice and oil palm cultivation in Southeast Asia, particularly in Indonesia. These climatic hazards affecting the main rice-producing regions (Java and Bali), increase the risk of annual production loss (Naylor *et al.* 2007) and encourage the development of innovative strategies to maintain stable production. Oil palm yields also decrease in regions where more and more drought periods are recorded (South Sumatra), and thus raise interest in limiting competition for water through adapted planting density. Subsequently, oil palm/rice intercropping in the rainy season alternating with palms alone at reduced density in the dry season could be a strategy to extend suitable area for rice while reducing competition for water between palms during dry spells. The feasibility of such a system depends on a specific management of the resources depending on the two species and seasons: the water availability for oil palm during the dry season and the transmitted light for rice during the rainy season. A main concern is therefore on adjusting the palm density to maintain high yields of both crops, and on screening rice varieties best adapted to shady conditions.

Research on oil palm agroforestry systems is rapidly expanding in the scientific literature (Bhagwat & Willis 2008; Khasanah *et al.* 2020). The majority of these studies focused on the ecological impact of these systems compared to conventional monoculture systems (Gérard *et al.* 2017; Ashraf *et al.* 2018; Zemp *et al.* 2019; Koussihouèdé *et al.* 2020; da Silva Maia *et al.* 2021). Rare studies highlighted the interest of palm agroforestry systems for their good agronomic performance and economic profitability on the one hand (Gawankar *et al.* 2018; Ahirwal *et al.* 2021), but also for buffering extreme climatic events and thus improving the sustainability of these systems (Ashraf *et al.* 2019). However, very few studies proposed intercropping with rainfed rice (Alridiwirsah *et al.* 2019), a crop that is nevertheless predominant in Asia. Oil palm-based intercropping is often considered at juvenile stage of oil palm, as a transitory option for a couple of years, since less than five percent of the incident light forages the canopy when oil palms reach maturity at a conventional density of 136 or 143 plants ha^{-1}. So far, no study investigated how changes in planting patterns and density of oil palm could be optimized for intercropping purposes, because i) such study requires years and strong

financial support, and ii) suitable planting management that optimizes palm productivity while providing enough light for undergrowth crop is not straightforward. The recent development of modelling tools, using 3D models of plants in combination with light modelling (Perez *et al.* 2022) now opens the way to test how planting design would modulate light availability under oil palm canopy without involving expensive trials.

Despite the growing interest in rice-based agroforestry systems (Wangpakapattanawong *et al.* 2017; Rodenburg *et al.* 2022; Masure *et al.* 2022), the diversity of rice response to fluctuations in radiation as well as low light remains unexploited. Up to now, varietal improvement programs in rice have focused on the selection of productive materials in full sun conditions. The existence of genetic variability in response to photosynthesis induction rate during shade/light alternations (Qu *et al.* 2016; Acevedo-Siaca *et al.* 2020; Taniyoshi *et al.* 2020), and independently to shade tolerance (Wang *et al.* 2015a; Dutta *et al.* 2018), suggest that it could be possible to select rice material suitable for agroforestry. Nevertheless, these studies investigated shade tolerance using continuous artificial light attenuation with shade nets, often starting at the reproductive stage of plant growth, thus preventing mimicking light conditions in agroforestry, where undercrop can be exposed to full sunlight around noon. In these systems, the light environment is much more complex than a constant light attenuation, since both effects of shading and daily light fluctuation can affect plant growth, from germination to harvesting.

This study raises the question of whether rice genotypes selected to be most efficient under optimal radiation are most efficient when the amount and the dynamics of radiation are altered, as found in agroforestry systems. Given the nonlinear relationship between light interception and photosynthesis, not only the reduction in light quantity but also the changes in daily light fluctuation can affect resource acquisition and subsequently plant growth, development and yield. Using a functional-structural plant model (FSPM) estimating light transmission through 3D canopy, we first simulated how palm density and planting design affect the quantity of transmitted light to rice, and selected a pattern optimizing light foraging for rice and oil palm density. Secondly, we reproduced in controlled conditions the specific changes in light intensity and fluctuation obtained in the simulation. Three light treatments differing in either the quantity or the dynamics of radiation were thus set up to investigate changes in phenological, morphological, physiological, and productive traits including physicochemical traits on eight contrasted rice varieties.

Material and Method

Modelling light transmission in innovative oil palm-based agroforestry system

We can test how planting design can modulate light availability under oil palm canopy by coupling a 3D structural model of oil palm (Perez *et al.* 2016, 2018) with a light interception model (Dauzat & Eroy 1997). This modelling approach helps to compute the light transmitted over time and space according to any design and thus allows the exploration of innovative and suitable planting patterns for agroforestry systems. In this study, we assumed that the architectural plasticity of oil palm had negligible impact on light capture in comparison to the changes in light interception due to planting patterns, as concluded in a previous study (Perez *et al.* 2022). A simulation study was performed with 3D mock-ups of oil palm representing plant architecture at the mature stage (8 years after planting), and displayed in rows defined by two parameters: the distance between rows and the distance between plants within rows (Fig S1). Eight planting patterns were designed *in silico* to estimate the light transmitted through the palms in a horizontal plane placed one meter above the ground (Table 1).

Table1: The eight designs tested with the corresponding distance between plants and density. The two parameters to define a design are the distance between rows (dist inter) and within rows (dist intra).

dist intra (m)	dist inter (m)	density (plants ha^{-1})
8	14	89
7	14	102
8	12	104
7	12	119
8	10	125
8	9	138
7	10	142
7	9	158

Distances within rows (7 and 8 m) were selected according to the average crown dimensions of mature oil palm, with the objective to mutual shading. Distance between rows (9, 10, 12, 14m) were selected to modulate planting density from low (89 plants ha^{-1}) to high density (158 plants ha^{-1}). Simulations were done using radiative data collected in South Sumatra on March 20th, 2019 (sunny day), with a

global radiation reaching 28.4 MJ m^{-2}. The light transmitted under the canopy was integrated over the day for every element of the horizontal plane (each element is a 16 cm x 16 cm square). Considering the space for growing rice at least 3 m distant from oil palm rows, we estimated on average photosynthetically active radiation (PAR) available in the rice area over a day.

Reproducing the simulated agroforestry light conditions in growth chambers

The experiment was carried out in the Abiophen platform of Montpellier in controlled conditions (Aralab Fitoclima 25000HP Growth chamber) in which light intensity and time course were adjusted while maintaining equivalent temperature and air humidity (Fig S2). Three light treatments (light supplied using Philips Master Colour CDM-T Elite MW 315W) were designed to test independently the effect of daily light quantity and the effect of diurnal fluctuations of light on rice. The three light treatments were set up in three independent growth chambers, respectively:

- a control treatment (C) following the dynamic of incident light observed in Sumatra (with a ceiling radiation value that can be reached *i.e* 1100 $\mu mol_{photons}$ m^{-2} s^{-1})
- the agroforestry shading regime corresponding to the planting design selected from the simulation study (S-AF)
- a uniform shading treatment (S)

The light reduction in the S treatment was performed to obtain the same amount of total daily radiation than in the S-AF treatment (Fig 1A), and corresponded to a constant shade that could be obtained with a shading net. We characterized this amount of light by integrating the photosynthetic photon flux density (PPFD) over a day, hereafter called daily light integral (DLI, Poorter et al. 2019). The DLI were 32.4 mol m^{-2} d^{-1}, 12.9 mol m^{-2} d^{-1} and 12.9 mol m^{-2} d^{-1} for C, S-AF and S, respectively, both shading treatments thus providing equivalently 60% of light attenuation (Fig 1B). Regarding light fluctuation, we calculated an index of daily light fluctuation (DLF) as the daily difference between the maximal and the minimal PPFD (i.e 0 $\mu mol.m^{-2}$ s^{-1}). DLF for S and S-AF were 478 and 694 μmol m^{-2} s^{-1}, thus representing respectively a reduction of 52% and 30% of light fluctuation compare to the control treatment. Light spectra were also compared in the three growth chambers to ensure similar light quality among treatments (Fig S3).

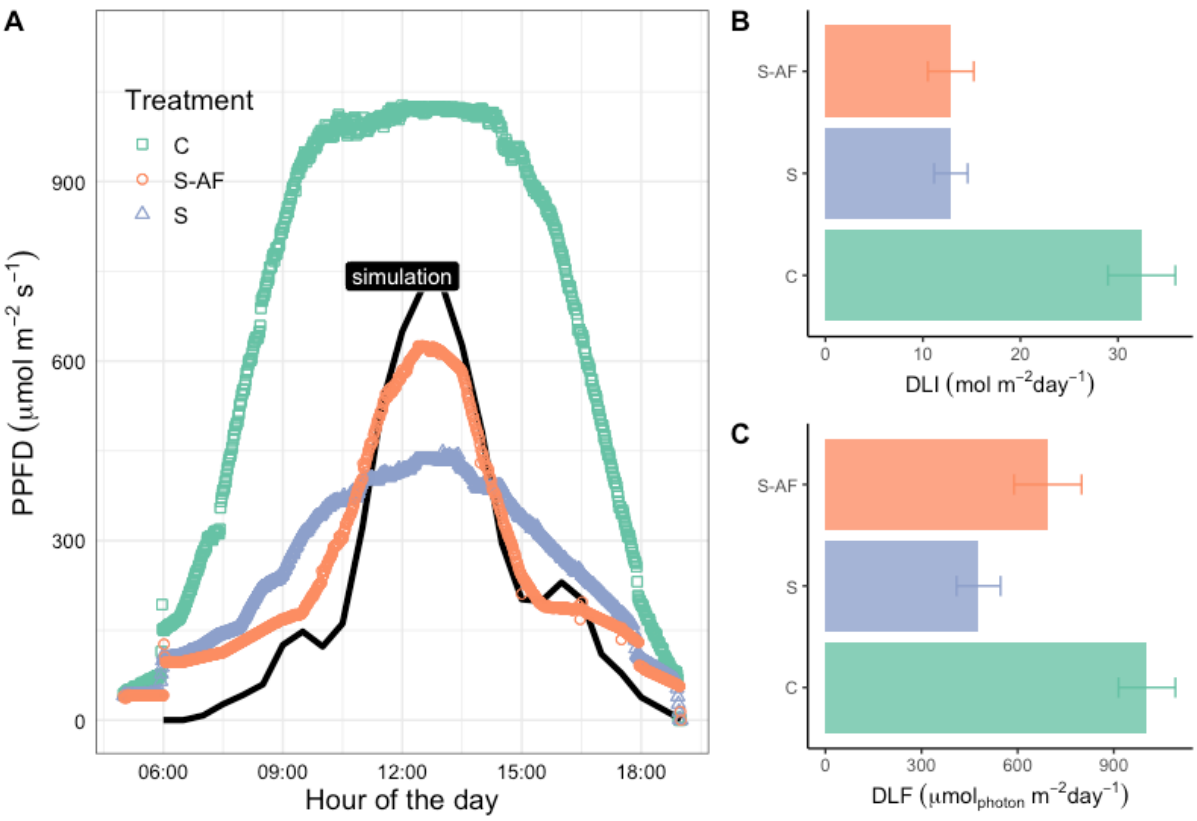

Figure 1: A) Mean diurnal time course of the photosynthetic photon flux density (PPFD) measured under three light treatments in the phytotrons during the whole cycle of plant growth (C: control. S-AF: agroforestry-like shade. S: constant shade). The black line indicates the PPFD simulated in silico in the 7m x 12m planting design. B) Daily light integral (mean and inter-days standard deviation) measured for the three treatments. S and S-AF represent 60% loss of radiations in comparison to C. C) Daily light fluctuation (mean and inter-days standard deviation) measured for the three treatments. S and S-AF represent respectively 52% and 30% loss of light intensity in comparison to C.

Plants material and phenotypic monitoring

Eight accessions of rice representing diversified subpopulations and geographical regions were provided by the seeds and genetic resources laboratory (TBRC) held in Montpellier, France (Table 2). Seeds were sown in individual pots of 4.8L (16cm x 16cm x 23cm) and fertilized using 3g L^{-1} of NPK fertilizer (50% of Basacote 13-5-18 and 50% of Bio 5-3-8; Compo – expert). In each of the three growth chambers, accessions were grouped in tables of 28 plants (corresponding to a density of 32 plants m^{-2}) to prevent any competition for light among varieties within the table. Eight tables were placed in each phytotron, and pots were kept flooded to prevent any water stress. Temperature and relative humidity followed the average value observed in South Sumatra: temperature varied from 22°C by night up to 32°C during the day and relative humidity from 83% down to 65% during the day (Fig S2). Temperature and vapour pressure deficit were monitored similarly in the three treatments so that phenotypic variations were interpreted as the consequence of light treatment alone.

*Table 2: Rice accessions monitored during the experiment in controlled conditions. *accessions with full phenotypic monitoring from sowing to harvesting.*

Accession	Ref	Gervex	Subpopulation	Origin
JC1	V1	9 091	Basmati	India
MALAGKIT PIRURUTONG	V6	8 182	Japonica	Philippines
IR64 *	V3	8 531	Indica	Philippines
TREMBESE *	V7	43 675	Japonica	Indonesia
SINNA SITHIRA KALI	V4	51 064	indica	Sri Lanka
IRAT 177 *	V8	7 415	Javanica	Guyane
BRASILEIRO *	V5	809	Japonica	Brasil
CASAMANCE V2	V2	1123	Glaberrima	Sénégal

Phenotypic monitoring was conducted from sowing to harvesting, with phenological, morphological, physiological and yield measurements (Table 3). On each table, the 10 inner plants among the 28 plants constituting the canopy were studied to prevent border effects in the analyses. Phenological observations (leaf tip appearance on main stem and tillers count) were made once a week. For leaf appearance a Haun scale was used: the leaf was counted once its tip was visible and the total number of emerged leaves was expressed on a decimal scale, with partially emerged leaves counted according to the visible fraction of the full size (Haun 1973; Tivet *et al.* 2001).

Table3: Phenotypic variables collected during the experiment

Type	Variable	Unit	Definition	# replicates (per genotype and treatment)	Scale	Method
phenology	DaysFL	days	Number of days from sowing to flag leaf appearance	6	plant	manual
	Phyllochron	days leaf^{-1}	Average period between the sequential emergence of leaves on the main stem			
Plant morphology & architecture	#Til		Maximal number of tillers			
	#Lea		Total number of leaves on main stem			
	SGA	cm^2	Stay green area			image analysis
	PH	cm	Plant height (from collar to flag leaf)			manual
	CPN	-	Plant compactness	4		image analysis
	PLA	cm^2	Plant leaf area at flowering			planimeter
	SLA	cm^2 g^{-1}	Specific leaf area of total green leaves at flowering	4		oven and scale
	VDW	g	Total aerial vegetative dry matter at harvesting	4		planimeter, oven and scale
Le af mo	Length	cm	Flag leaf length	10	Fla g lea	manual

	Width	cm	Flag leaf maximum width			
	Area	cm^2	Flag leaf area	4		planimeter
	Incli	degree	Flag leaf inclination			manual
	StoDen	$m\ m^{-2}$	Flag leaf stomatal density			microscope and image analysis
Ecophysiology	ETRm	$\mu mol_e m^{-2} s^{-1}$	maximum rate of electron transport	6		fluorimetry (MINIPAM)
	alpha	$\mu mol_e\ \mu mol_{photons}^{-1}$	initial slope of the light curve, related to maximum yield of photosynthesis			
	Ik	$\mu mol_{photons} m^{-2} s^{-1}$	PAR value of the point of intersection between the horizontal line ETRmax and the extrapolated initial slope			
	Asat	$\mu mol_{CO2} m^{-2} s^{-1}$	Saturating assimilation			gas exchange (Licor 6800)
	Aqe	$\mu mol_{CO2}\ \mu mol_{photons}^{-1}$	apparent quantum efficiency			
	Lcp	$\mu mol_{photons} m^{-2} s^{-1}$	Light compensation point			
Production	Yield	g	Plant grains dry matter	6	Plant	manual
	#Pan		Panicles number			
	#Kern		Number of grain per panicle		Panicle	
	TKW	g	1000 grains dry weight		Grain	
	SF	%	Spikelet fertility rate		Plant	
	HI	%	Harvest index			oven and scale
Post harvest quality	M	cm	kernel major axis (length)	3	grain	image analysis
	m	cm	kernel minor axis (width)	3	grain	image analysis
	Star %	$g.100g^{-1}$ db	flour starch content on dry basis	3	grain	enzymatic colorimetry
	Amy %	$g.100g^{-1}$ db	starch amylose content on dry basis	3	grain	differential scanning calorimetry
	ε %	%	pore volume to total volume ratio	3	grain	pycnometry and hydrostatic balance

At flowering, four out of the ten studied plants were collected for image-based analysis and destructive measurements. Destructive measurements consisted in total plant leaf area (PLA) using a planimeter and the shoot vegetative dry matter (leaves and stems). The ratio between PLA and the leaves dry matter gave an estimate of plant specific leaf area (SLA). The stomatal density (StoDen) was observed on flag leaves, from a footprint taken by a nail polish on the leaf surface and counted with ImageJ on the images observed on the microscope (Fig 2A). An index of plant compactness (CPN) was derived

from plant image, expressed as the ratio of the projected area of leaves over the projection of the volume defined by the whole aerial part of the plant (Fig 2B).

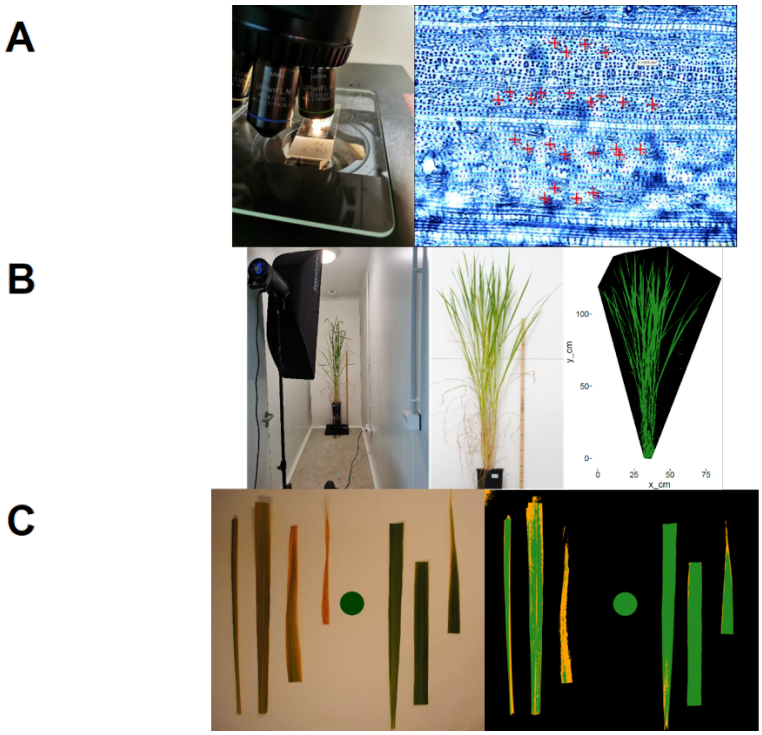

Figure 2: Image-based phenotypic variables. A) Right: Stomatal density observed on microscope slide, left: microscope image with stomata indicated with red crosses within a fixed area of 0.1 mm². B) Whole plant picture to retrieve compactness. Compactness is estimated as the ratio between the number of green pixels over the total pixels defining the convex hull of the areal part of the plant (green and black pixels). C) Estimation of the stay green area (SGA) from a picture of green leaves at harvesting. SGA corresponds to the area in green defined by green pixel. The green coin in the middle is used as a green reference for the segmentation process and scaling the pixel.

At harvesting, the six remaining plants were dissected into vegetative parts and reproductive parts. On the vegetative part, the total dry mass was weighted (vegetative dry weight: VDW) and pictures of the green leaves were used to estimate a stay green area (SGA) from pixel segmentation, defined as the green area left on the main stem (Fig2 C). On the reproductive part, the yield was measured as the total grains dry matter collected on all the panicle of the plant. The panicle of the main stem was used to count the number of fertile and unfertile grain per panicle and derive a spikelet fertility rate (SF). The harvest index (HI) was estimated as the ratio of the yield over the total aerial biomass (vegetative and panicles structural components).

Further analyses were performed on grain morphological and physicochemical feature on a subsample of the harvested spikelets. Paddy kernels were oven-dried at 45°C in a ventilated oven until constant weight was reached. Kernels were then dehusked and polished using a Satake model laboratory testing husker THU35C-T and Tadd mill 4E-230 machines. Digital RGB images (800 dpi) of polished kernels were acquired using a flatbed scanner (Epson Perfection v700 Photo). The grain major (length *M*) and minor axis (width *m*) were computed from the projected area of the binarized image of individual

kernels fitted from an ellipse, using ImageJ analysis software platform. Fine-ground rice flours were obtained using cargo kernels a ball mill (Dangoumill 300, Prolabo, France). After fine-group flour moisture content determination by thermogravimetry at 104°C, total starch content (%db) was estimated in triplicate by enzymatic colorimetry (515 *n*m) from the total released glucose after incubation and hydrolysis of about 300 mg db of flour with thermostable-α-amylase (Novo Nordisk, Denmark), and later with glucose oxidase and peroxidase (GOD & POD, Sigma, USA), as per Giraldo Toro et al. (2015). The trace of free glucose was neglected in the native rice flours. The amylose content was then estimated in duplicate using Differential Scanning Calorimetry DSC 8500 (Perkin Elmer, USA) as per Pérez et al. (2013) with modifications. The enthalpy change of formation of complexes between amylose and phospholipids during cooling was estimated from 10-11mg db of flour mixed to 40µl of lysophospholipid 2% (w/V in water) against a pure amylose standard. The kernel porosity (ε = 1- *app d*/d) expressed as percentage was then computed based on the estimation of the true density (*d* in $g.cm^{-3}$) of fine-ground flourby pycnometry using helium carrier and the apparent density (app d in $g.cm^{-3}$) using few kernels on a hydrostatic balance with ethanol. and expressed as percentage.

Gas exchange measurements were conducted on the flag leaf few days after flag leaf emission on the six plants monitored until harvesting. Response curves to photosynthetic photon flux density (PPFD) were performed using a portable photosynthesis system (LI-6800; LiCor Biosciences, Lincoln, NE, USA) at around 10 a.m. each morning over a week. Assimilation- irradiance (A – PPFD) curve fitting analysis were then conducted using the Mitscherlich function (Eq 1) (Heschel *et al.* 2004) to estimate the three photosynthetic parameters: the light saturated assimilation (A_{sat}), apparent quantum yield (Aqe) and the light compensation point (L_{CP}) for each individual.

$$A = A_{sat}\left[1 - e^{-Aqe(PPFD - L_{CP})}\right] \qquad \text{(Eq1)}$$

Electron transport rate - irradiance (ETR-PPFD) curve fitting analysis were also determined using equation (Eq 2) to extract the maximum electron transport rate (ETR_m), the quantum efficiency of photosynthesis (α) and the minimum saturating irradiance (I_k=ETR_m/ α).

$$ETR = ETR_m \tanh(\alpha\ PPFD/ETR_m) \qquad \text{(Eq2)}$$

Assessing the effect of daily light integral (DLI) and daily light fluctuation (DLF) on phenotypic traits

The three treatments considered in this study varied either from the amount of light receive (DLI) and/or in the intensity of light fluctuation perceived over a day (DLF). Each light treatment was a single combination of DLI and DLF. To evaluate if the difference in phenotypic variations were due to DLI or DLF, we conducted statistical analysis based on analysis of covariance models (ANCOVA). For each phenotypic variable studied, we tested the two following ANCOVA models:

$$Model\ 1: Y_{ij} = \mu + V_i + DLI + DLF + V_iDLI + V_iDLF + \varepsilon_{ij} \quad \text{(Eq 3)}$$

$$Model\ 2: Y_{ij} = \mu + V_i + DLI + V_iDLI + \varepsilon_{ij} \quad \text{(Eq 4)}$$

Where Y_{ij} is the phenotypic value of the individual j of the variety i; μ is model grand mean; V_i is the effect of the variety I; DLI and DLF are covariates, V_iDLF and V_iDLI are the interaction terms and ε_{ij} is the residual term. Normality of model residuals and homoscedasticity were tested, and response variables were transformed using square-root or Box-Cox transformation accordingly to ensure normality. For each phenotypic trait, we conducted model testing by comparing the two models fit using a likelihood ratio test (LRT):

$$LRT = -2(\log LK_{model1} - logLK_{model2})$$

Were logLK is the model log likelihood. LRT was compared with a Chi-squared distribution to obtain a p-value. Significant differences between the two nested models (p-value<0.05) meant that DLI alone did not explained the phenotypic variations observed, and therefore that considering the changes in light fluctuation (DLF) improved model prediction. Likelihood ratio tests were performed using the lrtest function of the lmtest R package (Zeileis & Hothorn 2002).

Results

Defining the light regime available for growing rice under an oil palm-based agroforestry system

The simulation of light transmission on the eight planting patterns investigated clearly demonstrated the significant reduction in the light transmitted with increasing planting density (Fig 3). For the lowest density, light transmission reached a maximum of 43% in the middle of the inter row, while it only reached 13% for the highest density. Simulation outputs showed an important decrease in radiation with increasing palm density, with a maximum irradiance around midday. The daily radiation available

for rice decreased with planting density (Fig 4), the selection of an interesting design hence depended on the trade-off between oil palm density and daily PAR transmitted for rice.

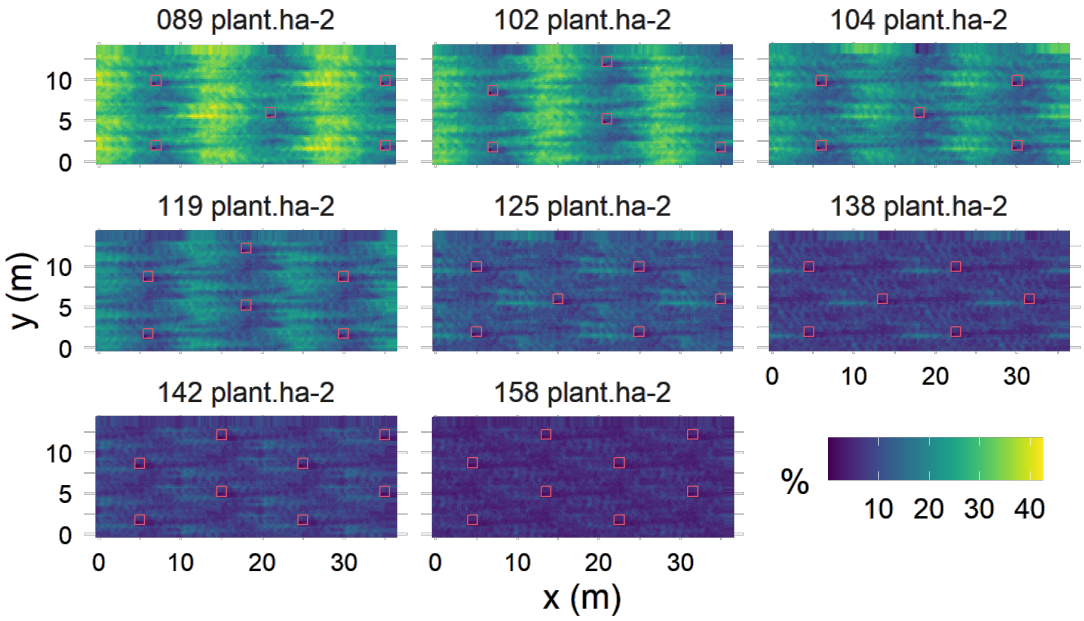

Figure 3: Simulated light transmission under oil palm canopies with various planting designs. Colors indicates percentage of incident light and red squares the positions of palm trees.

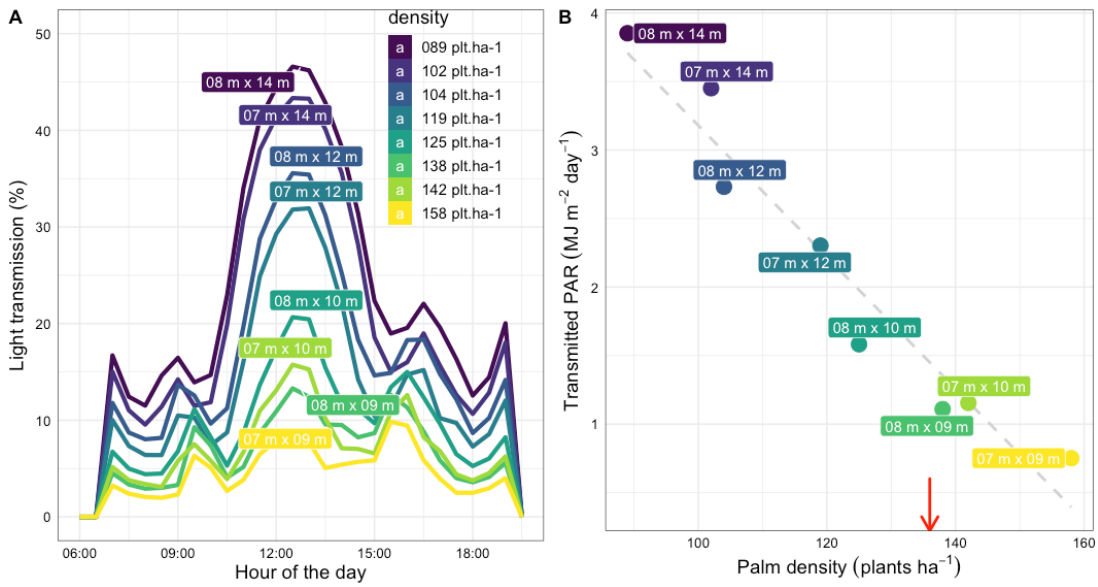

Figure 4: A) Average light transmission in the rice area over a day for the height simulated designs. B) Average daily photosynthetic active radiation available for rice between palm rows. The red arrow represents the conventional density (136 plants ha^{-1}) and the dotted line the average loss of transmitted light with increasing density.

In one hand, the distance between rows strongly influenced the intensity of light around midday (Fig 4A), distances lower than 12m sharply limiting the maximum light transmission below 20%. In the other

276 hand, distances over 12m decreased oil palm density below 110 plants ha^{-1}. The 7 m x 12 m design
277 exhibited the best compromise to ensure a sufficient palm density and limiting shading between rows.
278 Moreover, this density of 119 plant ha^{-1} is close to the conventional densities of 136-143 plants ha^{-1}.
279 The 7 m x 12 m design was thus chosen as the reference to mimic in controlled conditions rice-palm
280 agroforestry light, i.e the 60% of light attenuation over the day (Fig 1).

281

282 *Changes in plant morphology are mainly explained by daily light quantity while*
283 *plant phenology is also impacted by light fluctuation*

284 Under the shading treatments two varieties showed drastic increased in leaf length and etiolation
285 symptoms leading to severe lodge, which prevented conducting them up to harvest: varieties V1 and
286 V2 were thus removed from the three growth chambers 55 days after sowing. This first result pointed
287 out how morphological responses to shade could introduce additional yield-reducing factor as lodging,
288 and can be a first criterion to select adapted varieties. The variety V6 did not reach production maturity
289 within the experimental time window, and the variety V4 was sensitive to the photoperiod chosen for
290 the experiment, thus preventing plants to flower. Consequently, the following results will focus on the
291 four other varieties on which phenotypic monitoring was performed from sowing to harvest.

292 Reduction in daily light quantity significantly affected the dynamic of leaf and tiller emission (Table 3
293 and Fig 5A & B). The phyllochron on the main stem increased under shade, with variations in the total
294 number of leaves emitted depending on treatments and varieties (Table 4). During the first month
295 after sowing, the number of leaves on the main stem was higher under the control treatment than
296 under shade, but at maturity, plants under shade produced more leaves on the main stem for the
297 *japonica* varieties (Fig 5A). The *indica* variety V3) emitted more leaves on the main stem under the
298 control treatment than in the S treatment whereas the japonica V5 presented an opposite behavior
299 with an increased number of leaves both shading treatments (Table 4). A significant effect of light
300 fluctuation was revealed on phyllochron and the total number of leaves on the main stem, with
301 important variations among varieties (Table 4). Regarding tiller emission, reduction in DLI induced
302 significantly less tillers. DLF modulated the total number of tillers emitted, with more tillers in S-AF
303 than in S, with significant differences for V5 and V8. Leaves length and width on the main stem sharply
304 increased under both shading treatment (Fig 5C), and these changes in leaf dimensions were attributed
305 to DLI and not to DLF (Table 3). The flag leaf shape (LW ratio) changed significantly for variety V3 with
306 a higher length to width ratio under shade, while the other three varieties maintained a comparable
307 shape and even tended to reduce the ratio by contrast (Table 4).

308

309 *Table 3: Summary of models analysis to assess the effect of variety, daily light integral (DLI), daily light*
310 *fluctuation (DLF) and their interaction on phenotypic traits. An ANCOVA is performed with variety as factor and*
311 *DLI and DLF as covariables (ns: non-significant, *P<0.05;**P<0.01;***P<0.001). For variable abbreviations, see*
312 *Table 2.*

	Rep	var	Variety	DLI	DLF	V x DLI	V x DLF	r2	LRT
Yield components	6	Yield	***	***	ns	ns	*	0.76	***
	6	#Pan	***	***	ns	ns	ns	0.83	ns
	6	#Kern	***	ns	**	ns	*	0.65	***
	6	TKW	***	***	***	**	**	0.89	***
	6	SF	***	*	***	ns	**	0.49	***
	6	HI	***	***	***	**	***	0.84	***
Flag leaf morphology	10	Length	***	***	ns	ns	ns	0.52	ns
	10	Width	***	***	ns	**	ns	0.77	ns
	10	LWratio	***	ns	ns	*	*	0.56	ns
	10	Area	***	***	ns	ns	ns	0.56	ns
	4	Incli	***	**	ns	ns	***	0.92	***
	4	StoDen	***	***	ns	ns	ns	0.73	ns
Phenology	10	DaysFL	***	***	*	ns	ns	0.66	ns
	10	Phyllochron	***	***	*	***	*	0.65	**
Plant architecture	10	#Lea	***	*	ns	***	***	0.77	***
	10	#Til	***	***	***	**	ns	0.94	***
	4	PLA	***	***	ns	ns	ns	0.47	ns
	4	CPN	***	***	ns	*	ns	0.84	ns
	4	SLA	***	***	ns	ns	ns	0.64	ns
	6	VDW	***	***	*	ns	**	0.54	***
	6	PH	***	***	ns	ns	ns	0.85	ns
Ecophysiology	6	SGI	***	***	***	***	**	0.90	***
	6	Lcp	***	**	*	*	***	0.61	***
	6	Asat	*	ns	**	ns	***	0.34	***
	6	Aqe	***	ns	ns	ns	***	0.48	***
	6	Rd	*	ns	*	ns	**	0.27	***
	6	ETRm	ns	ns	*	ns	*	0.21	**
	6	alpha	***	*	ns	ns	***	0.49	***
	6	Ik	**	ns	**	ns	ns	0.24	*
Grain morphology	3	M	***	ns	ns	ns	ns	0.95	ns
& physicochemical quality	3	m	***	ns	ns	ns	ns	0.48	ns
	3	Star %	*	**	*	ns	ns	0.49	**
	3	Amy %	***	ns	ns	ns	ns	0.97	ns
	3	ε %	***	*	***	***	***	0.92	***

313

314 Interestingly, stomatal density on flag leaf significantly decreased with DLI, but only for the two
315 varieties V7 and V8, while the two others varieties V3 and V5 kept a comparable stomatal density while
316 increasing flag leaf area. At the plant level, the decrease in DLI involved taller and less erected plants
317 (lower CPN), mainly due to longer, wider and thinner leaves as well as longer internodes (Fig S4).

Table 4: Phenotypic variations of traits with varieties and treatments. Numbers indicate mean value and letters represent significant difference between treatments for each variety assessed by Tuckey's test (P<0.05)

	V3			V5			V7			V8		
	C	S	S-AF	C	S	S-AF	C	S	S-AF	C	S	S-AF
Yield	34.2a	25.9ab	22.3b	38.7a	25.8b	31.2b	18.5a	14.1a	12.9a	33.4a	28.5ab	23b
Nb_Pan	14.8a	12.3b	13ab	9.33a	7.33a	7.67a	6.5a	5.17a	4.83a	10a	7.67b	8.33ab
Nb_Kern	92.1a	89.9a	73.7a	143a	138a	143a	115a	116a	111a	97.7b	**120a**	**89.1b**
TKW	27.9a	25.1a	27.9a	30.7a	**27.2b**	**30.7a**	27.1a	24.4b	26ab	36a	32.8b	32.2b
SF	85.9a	**83.2a**	**61b**	89.5a	86.4a	85a	85.8ab	**94.3a**	**76.1b**	94.2a	92.6a	90.2a
HI	35.2a	**37.3a**	**25.2b**	46.3a	**41.2c**	**43.4b**	26a	25.3a	23a	35.7b	**38.1a**	**33.8b**
DaysFL	77.5b	86.8a	85.2a	76.3b	85.7a	83.3a	74b	81a	81a	76.3b	82.2a	78.7ab
phylochron	5.11c	**6.28a**	**5.68b**	5.87a	5.98a	6.03a	5.93b	6.49a	6.32a	5.8a	5.94a	5.9a
Nb_Lea	15.2a	**13.8b**	**15a**	13b	14.3a	13.8a	12.5a	12.5a	12.8a	13.2a	13.8a	13.3a
Nb_Til	35a	22b	24.2b	19.3a	**10.5c**	**13b**	11.5a	7.33b	7.67b	23.7a	**11c**	**14.2b**
CPN	0.32a	0.24b	0.21b	0.24a	**0.12c**	**0.16b**	0.19a	0.14b	0.14b	0.19a	0.14b	0.14b
PH	88.8a	89.8a	97.5a	113c	**121b**	**128a**	137a	141a	148a	110b	123a	117ab
VDW	60a	**41b**	**63.7a**	43a	35.6a	38.9a	51.9a	40.4a	42.3a	58.2a	45.1b	44.3b
PLA	2316b	3279ab	3348a	2173a	2223a	2487a	1615b	2117ab	2515a	2063a	2552a	2630a
SLA	209b	252a	263a	177b	218a	220a	158b	229a	216a	165a	210a	215a
FL Length	31.4b	41.5a	39.2a	38.1b	47.2a	40.9ab	32.9b	41.6a	43.7a	46.3b	54.6a	56.7a
FL Width	1.33b	1.54a	1.48a	1.39b	1.79a	1.76a	1.69b	2.15a	2.12a	1.54b	1.94a	1.83a
FL LWratio	23.7b	26.9ab	26.4a	27.6a	26.5a	23.3a	19.4a	19.2a	20.6a	29.9a	28a	31a
FL Area	30.4b	47.6a	44.8a	40.5b	60.8a	50.7ab	37.8b	60.4a	59.5a	48b	75.2a	73.4a
FL Incli	168a	165a	166a	164a	161a	162a	93.8ab	**107a**	**63.2b**	160a	**128b**	**146a**
FL StoDen	748a	698a	671a	641a	623a	649a	594a	531ab	516b	583a	512b	544ab
SGA	70.8b	172a	120ab	3.66c	**176a**	**68.9b**	68b	344a	305a	16.2c	**165a**	**119b**
SPAD	31.9a	35.4a	36.3a	35.4b	**43.1a**	**36.7b**	41.6a	41.2a	35.3a	36.2a	36.1a	36.2a
Lcp	54.7ab	**59.5a**	**33.1b**	34.2a	22.3a	31.2a	39.6a	**29a**	**16.7b**	30.8a	32.4a	35.7a
Asat	16.6a	17.1a	19.8a	13.7b	**19.7a**	**11.7b**	17.2a	15.1ab	11.6b	18.1a	17.6a	15.4a
Aqe	0.0021b	**0.0021b**	**0.0028a**	0.0027ab	**0.0033a**	**0.0023b**	0.0029b	**0.0028b**	**0.0042a**	0.0031a	0.0029a	0.0027a
Rd	0.5a	0.63a	0.85a	0.57ab	**0.67a**	**0.25b**	0.75a	0.64a	0.34a	0.9a	0.85a	0.63a
ETRm	154a	146a	161a	153ab	**173a**	**136b**	163a	149ab	121b	168a	159a	152a
alpha	0.29ab	**0.27b**	**0.32a**	0.32a	**0.33a**	**0.3b**	0.33ab	**0.32b**	**0.34a**	0.33a	0.32a	0.31a
Ik	522a	546a	501a	476a	526a	458a	491a	**469a**	**356b**	505a	500a	490a
M	6.71a	6.67a	6.68a	6.22a	6.55a	6.33a	4.73a	4.99a	4.67a	6.79a	6.57a	6.65a
m	2.35a	2.25a	2.25a	2.43a	2.55a	2.44a	2.70a	2.67a	2.69a	2.60a	2.54a	2.55a
Star %	77.5a	77.2a	78.4a	84.5a	72b	79.2ab	84.6a	77.6a	75.4a	76a	68.6a	77.8a
Amy %	23.5a	25a	23.7a	14.2a	14.5a	14.8a	21a	21.7a	21.6a	12.8a	12.6a	10.6a
ε %	10.3a	9.60a	9.78a	8.45b	**10.7a**	**8.38b**	11.7a	**9.54c**	**10.3b**	8.66ab	**8.82a**	**8.29b**

The total plant leaf area increased under shading treatments without marked difference between S and S-AF. The SLA increased in both shading treatments without marked difference du to DLF, traducing the effect of light quantity alone in leaf elongation and widening. The effect of DLF was only important for the total vegetative dry matter (VDW) at harvesting, partly because of the contrasted behavior of V3 between S and S-AF (Table 4).

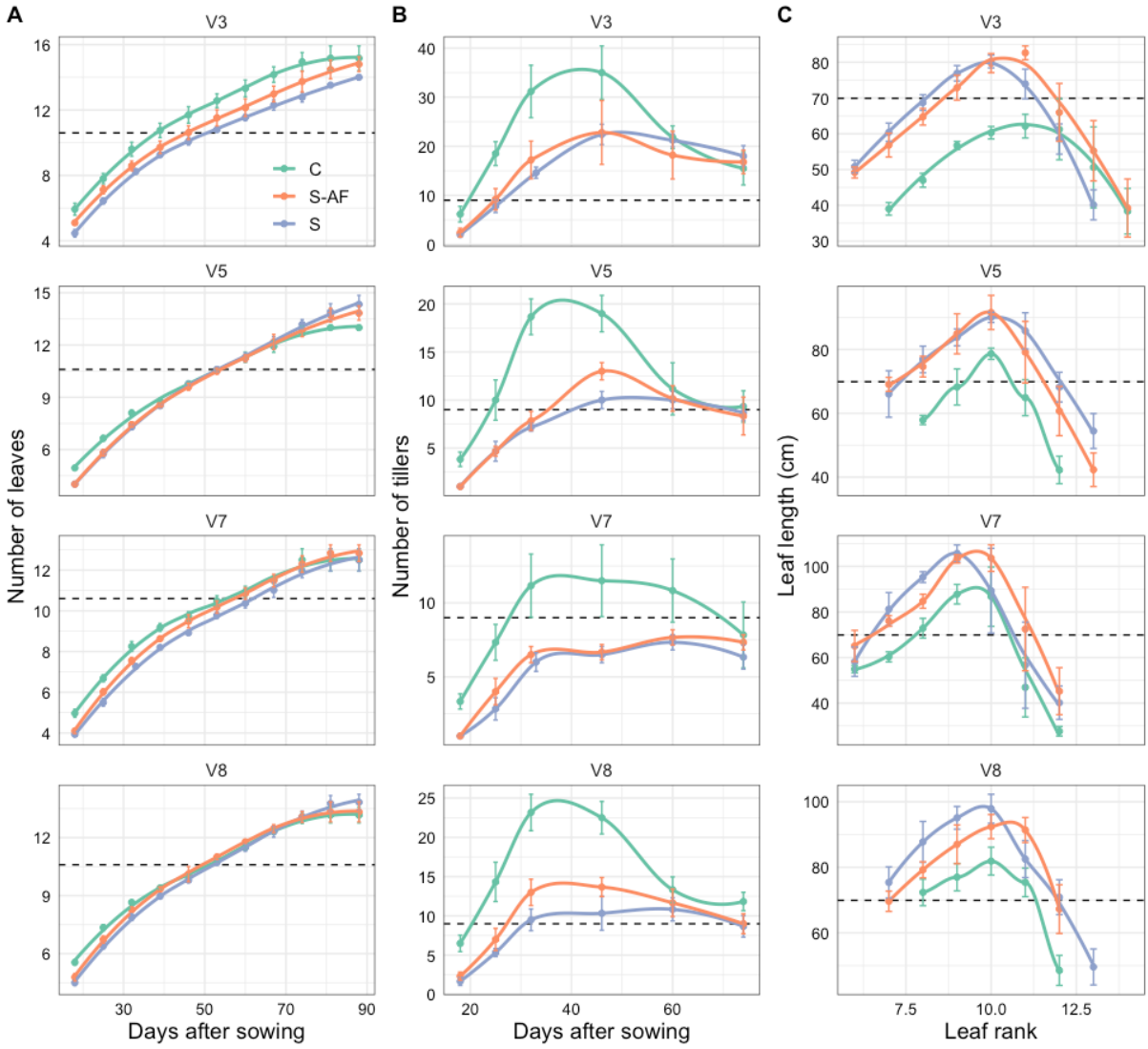

Figure 5: Dynamic of leaf emission according to varieties and treatments (A), dynamic of tillers emission according to varieties and treatments, (B) and length of leaves on the main stem (C). Smooth lines and points represent average values and vertical lines standard deviations. Dotted lines indicates the median value (taking all treatments and varieties together) for better visualize the difference between varieties.

Light fluctuation impacted leaf senescence and grain filling, and involved high variation in photosynthetic capacities among varieties

At the plant scale, the most marked difference between treatments was the stay green area at physiological maturity (Table 3 & Table 4). We observe a gradient in leaf senescence with light fluctuation, plants from the control treatment presenting higher senescence than plants under shade

that stayed greener. Increasing DLF significantly increased senescence (low SGA), suggesting that resource reallocation may depend on irradiance intensity.

Plant yield was significantly higher in the control treatment, and no major differences were stressed between the two shading treatments (Tables 3 & 4). Lower yield under shade was mainly explained by the reduction in the number of panicles per plant, but the establishment of grain yield varied with DLF. Indeed, the comparison of production between S and S-AF revealed that under higher light irradiance (S-AF), spikelet fertility was reduced, thus limiting the number of kernels per panicle. Nevertheless, yield was maintained under S-AF trough higher TKW (bigger grains) (Table 3), suggesting that grain filling was enhanced with irradiance intensity. The varieties V3, V5 and V8 produced more than the variety V7 in all the treatments. Interactions between varieties and shade treatments was stressed, the variety V5 producing significantly more than V3 and V8 under S-AF conditions while no significant difference was stressed between these three varieties in the two other treatments (Fig 6).

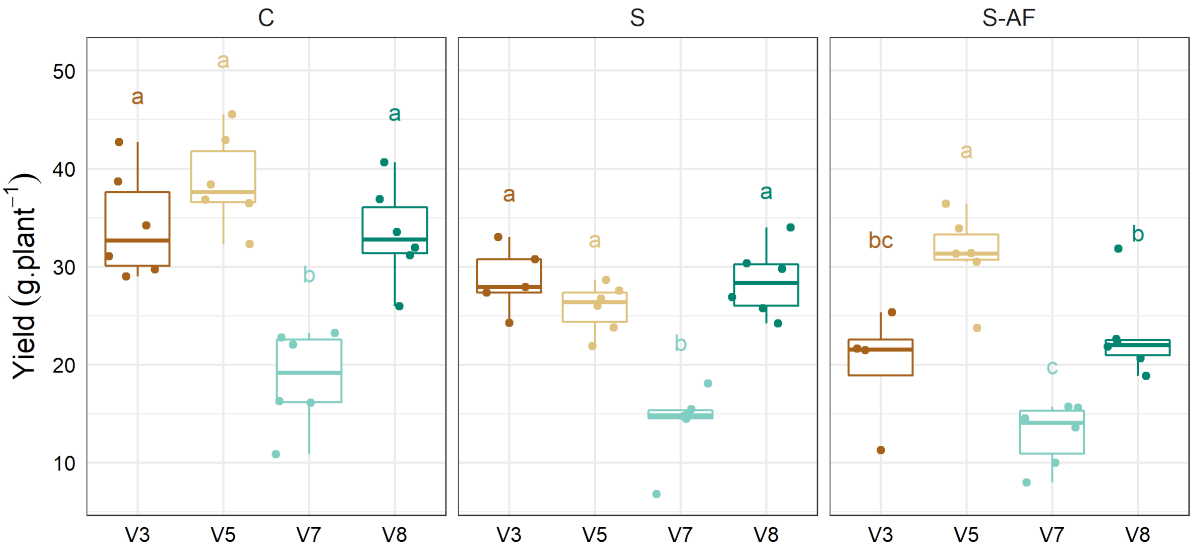

Figure 6: Plant yield for the four varieties studied in the three light treatments. Letters indicate significant differences between varieties in each treatment (Tuckey s test P<0.05).

The study of flag leaf CO^2 assimilation revealed high interaction effects (V x DLF) for nearly all photosynthesis parameters (Table 3). Light-responses curves exhibited contrasted behavior of varieties, with more varietal variability under the S-AF treatment than in the two other treatments (Fig 7). The variety with maximal photosynthesis was different in each treatment, revealing the high variability in photosynthetic processes with environment and the potential weak heritability of photosynthetic traits. No clear relationship could be established between photosynthetic capacity and other morphological or yield-related variables, highlighting the complexity and multifactorial

processes involved in plant growth and production. At the grain scale, the morphology of grain (length and width) was not significantly affected by the light treatments (Table 4). Complementary ANCOVA confirmed that DLI and DLF has no significant effect on grain morphology (Table 3) contrarily to the factor variety, suggesting that kernel length and width are heritable traits.

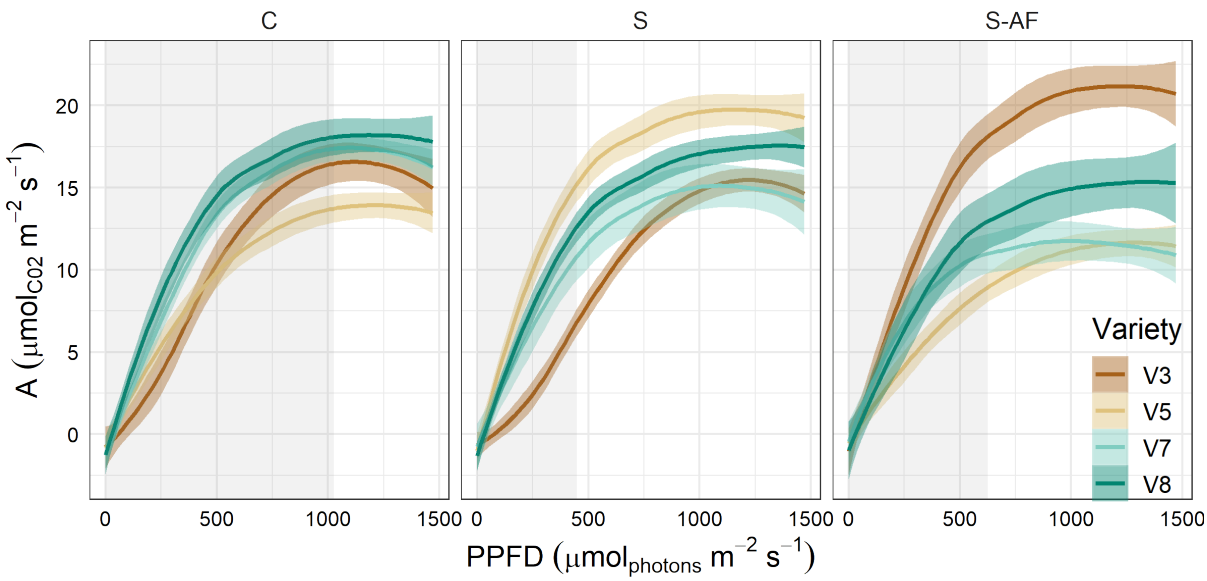

Figure 7: Response curves of carbon assimilation with irradiance according to treatments and varieties. Lines indicate smooth curves adjusted per variety and per treatment with 95% confidence interval. Grey areas represent the light fluctuation domain provided in each treatment.

Both amylose and starch content were revealed being also significantly affected among varieties. The variation of the composition of starch was only observed among V5 under low light, where the DLI treatment induced a significant reduction of the amylose content as compared to the control one. Varieties V3, V5, V8 exhibited significant variation in their porosity. The kernel porosity was significantly affected by varieties, by treatments and their interactions (Table 4). Contrary to V5 and V8 varieties, V7 presented a significantly lower level of air space for the kernel grown under S than that of S-AF environment.

Relative changes among varieties are maintained between the light treatments

Three principal component analysis (PCA) were performed, one for each treatment (Fig 8). For all PCA, more than 60 % of the phenotypic variance was explained by the two first components (PC1 & PC2), the contribution of PC3 to total variance being markedly lower than PC2. In the control treatment (Fig

8A), vegetative components (number of tillers, number of leaves) were the major contributors to PC1 while yield components (kernel weight, harvest index) contributed to PC2.

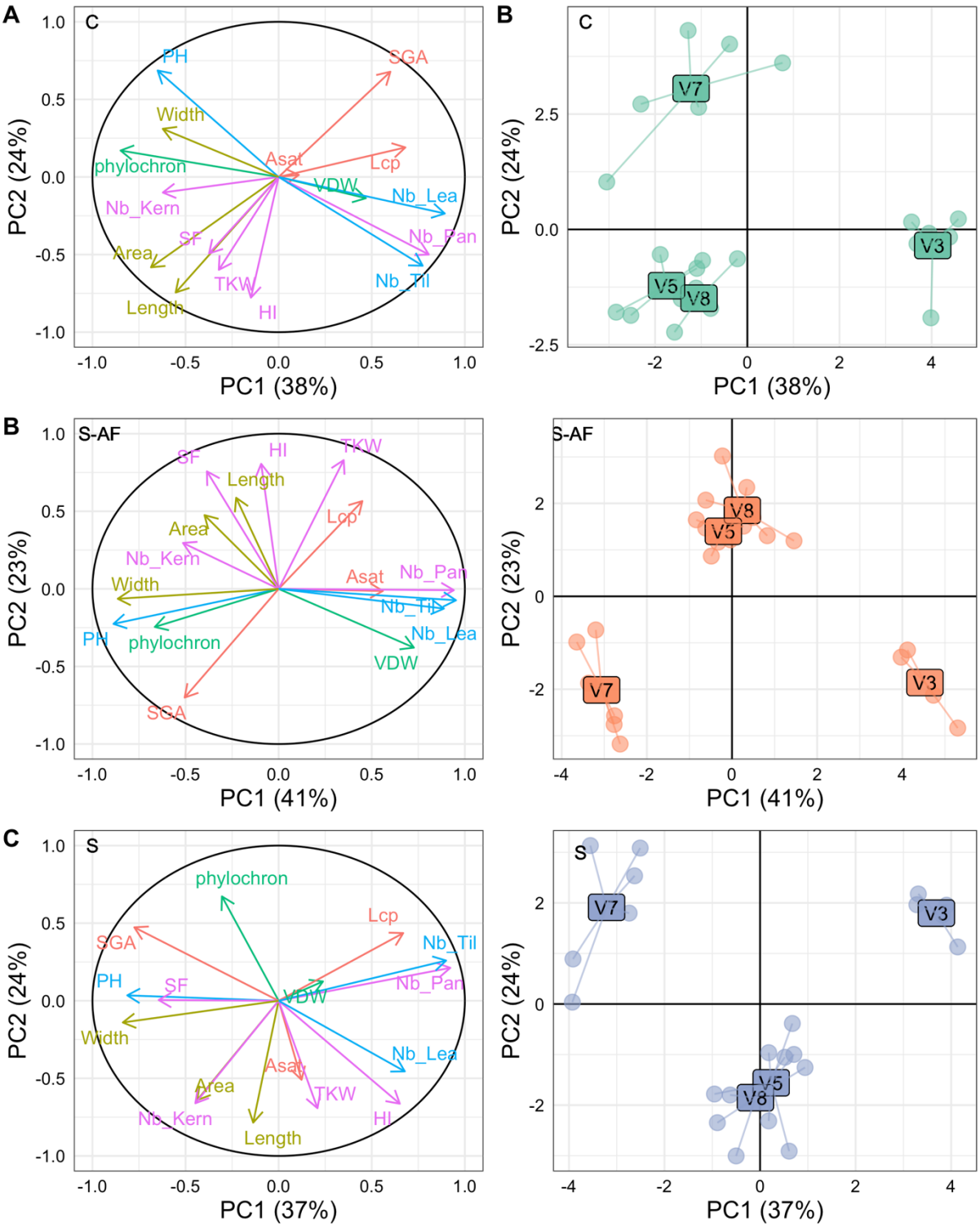

Fig 8: Principal component analyses performed on the three treatments (A: treatment C, B: treatment S-AF and C: treatment S). Graphics on the left represent the projection of traits on PCA axes and graphics on the right the projection on PCA of individuals grouped per varieties

387 Interestingly, flag leaf morphology (length and area) contributed more to PC2 than PC1, suggesting a
388 closer link between flag leaf geometry and yield components than between the whole plant vegetative
389 attributes and yield. PCA for S-AF (Fig 8B) resulted in similar interpretations, the vegetative
390 components contributing to PC1 and yield components to PC2. Additionally, plant height and flag leaf
391 width contributed markedly to PC1, which revealed the architectural variations triggered by lower light
392 intensity. Physiological variables such as Lcp, Aqe and SGA were highly correlated to yield components.
393 Plants with the highest yield presented low SGA (high senescence), low photosynthetic efficiency and
394 high light compensation point. This result confirmed that under S-AF treatment, yield depended more
395 on resource remobilization to the panicle rather than enhancement of the photosynthetic capacities
396 of the flag leaf. The PCA performed on the treatment S (Fig 8C) indicated only small discrepancies with
397 the PCA for the S-AF. The dichotomy between vegetative/architectural features and yield features was
398 marked in both shading treatments. Variations in flag leaf width particularly discriminated individuals
399 in shading treatments, and in contrast to the control treatment, a higher correlation was observed
400 between flag leaf with and yield than between flag leaf length and yield (Fig S5). Projection of
401 individuals on the three PCAs clearly split varieties, revealing that the studied phenotypic traits and
402 their combination were specific to varieties. In the three treatments, projections of V5 and V8 were
403 superposed, denoting the phenotypic similarities between those two varieties. Architectural features
404 of V3 and V7 were significantly different since there projections were in opposition in respect to PC1,
405 notably in the shading treatments. Finally, the similarity between varieties projections from the three
406 PCAs may suggest that varietal responses to low light are likely to be heritable, and that varietal
407 screening under full light can provide clue on varietal behavior under low light.

408 Discussion

409 *Using functional-structural plant models to design innovative agroforestry*
410 *systems*

411 The development of agroforestry system often depends on the success of smallholders' initiatives to
412 test new farming practices. The adoption of oil palm agroforestry systems has been promoted among
413 smallhoders through empirical practices conferring socio-economic and environmental benefits
414 (Susanti *et al.* 2020). However, it is not straightforward to conduct specific studies to understand and
415 optimize the agronomic performance of oil palm agroforestry systems, since such studies require years
416 of investigation and are very costly. Intercropping is generally only associated in the first years after
417 planting, in systems where the density of oil palm is optimized for growing the palm alone at maturity
418 (Koussihouèdé *et al.* 2020). In this study, we considered that future food security and environmental

concerns would require the development of sustainable oil palm agroforestry systems, with the possibility of growing food crop in association with mature oil palms.

In this context, we proposed a new approach, based on the assessment of light distribution in a mature oil palm agroforestry system. Using a quantitative trade-off between light resource for rice cultivation and oil palm density to optimize oil production, the modelling approach proposed in this paper showed how FSPM could help designing oil palm-based intercropping systems with scientific background. The proposed methodology could be adapted to many systems that requires time-consuming and expensive trials. However, it is worthy to note that this approach was particularly suitable for oil palm-based systems due to the low architectural plasticity of the oil palm, and its low impact on light interception efficiency at the plot level (Perez *et al.* 2022). The low plasticity of oil palm reinforces its suitability for association with other crops. In this study we did not consider the effect of planting density on oil palm architecture and subsequently light interception, but it is likely that the same approach with perennial plants with complex branching patterns and highly plastic to the radiative environment would require additional modeling inputs to derive reliable light partitioning predictions. The number of designs tested in the presented simulation study was limited to a few numbers, but further simulations could be performed to mathematically optimize light partitioning and oil palm density, including new patterns different from the row planting proposed here.

Studies focusing on the effect of light attenuation are often performed in outdoor conditions using shading net (Liu *et al.* 2014; Chen *et al.* 2019; Jia *et al.* 2021), and more recently in growth chambers (Yang *et al.* 2018; Zhang *et al.* 2020). Highly controlled growth chambers are generally used to simulate future climate conditions (i.e rising CO2 or temperature), but can now be used to simulate change in light regime (Vialet-Chabrand *et al.* 2017). The originality of our study was to use modeling outputs to set up light treatments in controlled conditions. Conversely to outdoor conditions where the dynamic of light cannot be controlled, here we were able to investigate the impact of both light quantity and the dynamic of light intensity. Such experimental device aimed at mimicking the light regime of an agroforestry system not yet implemented, taking into account the reduction of light and its variation over the day. However, the major drawback of our experiment was the inability to modulate light quality (red/far red ratio) as it would be in the shade of oil palm, which can affect growth and development of the understorey crops (Chen *et al.* 2019; Zhang *et al.* 2020).

Toward the selection of rice varieties adapted to agroforestry light conditions

Three characteristics of light are modulated under agroforestry systems: the quantity of light, the fluctuation of light, and the quality of light. The inexistence of real conditions to screen the tolerance

452 of rice varieties to the agroforestry system of interest involved setting up a trial in controlled conditions
453 mimicking these conditions. Our controlled conditions prevented us from modulating light quality, and
454 more precisely the red to far-red ratio, which is known to be a light signal triggering a suite of
455 physiological and morphological responses called the shade avoidance syndrome (Franklin 2008;
456 Ballaré & Pierik 2017). Zhen and Bugbee (Zhen & Bugbee 2020) recently stressed the importance of
457 far-red photons on plant photosynthesis, suggesting the need to integrate a larger range of wavelength
458 in studies dedicated to investigate responses to light. Other studies already demonstrated the genetic
459 inheritance of cereal crops in responses light quality (Wille *et al.* 2017), but only few investigated the
460 interaction between light quantity and quality (Yang *et al.* 2018). Further dedicated experiments
461 integrating response to light quality, in combination with light quantity and light fluctuation would
462 thus be of great interest to complement the result found in the present study and better understand
463 responses to agroforestry light regime. However, in the intercropping design prospected here, change
464 in light quality would appear only a few hours during the day. This part time change in light quality
465 raises new question about how long the light signal must last to trigger the shade avoidance syndrome
466 (Li *et al.* 2021).

467 In our experiments, we focused on the effect of light integral (DLI) and light fluctuation (DLF) on rice
468 growth and production. DLI represented an index of resource availability, whereas DLF was considered
469 as a light signal. Our results confirmed that the reduction in the quantity of light on rice prolonged the
470 period of growth, increased plant height and leaf area, decreased the number of tillers and total plant
471 dry mater (Liu *et al.* 2014) and subsequently reduced grain yield (Wang et al. 2015b; Chen et al. 2019).
472 Our result clearly showed the changes in resource investment under low light: plants produced fewer
473 tillers but increase the number and the size of leaves in the main stem. Shading treatments triggered
474 leaf elongation and widening, conferring a higher leaf area than under high light intensity. In parallel,
475 leaves were thinner under shade, traducing a lower capacity to accumulate nitrogen (Murchie et al.
476 2005). The adjustment in leaf morphology were consistent with other studies in rice (Liu *et al.* 2014)
477 or others crops (Arenas-Corraliza *et al.* 2019; Poorter *et al.* 2019), notably the increase of SLA, but
478 showed some discrepancies with respect to leaf widening. Contrarily to the present study and the
479 review by Liu (Liu *et al.* 2014), reductions in leaf width under shade were observed for rice (Lafarge *et*
480 *al.* 2010) and for maize (Lacube *et al.* 2017). These discrepencies could be explained by the duration
481 and the intensity of shade, and raises question about the existence of light signal thresholds that
482 trigger morphological adjustments.

483 Light quantity was the main factor driving morphological and architectural changes, while light
484 fluctuations only appeared to explain variations in yield components and phenology (Table 3). This
485 study highlighted the importance of light fluctuation in the grain filling process and resource

486 reallocation. The stay green area was particularly affected by the intensity of light experienced by the
487 plant, revealing that fluctuating light matters to trigger some process linked to resource reallocation
488 and subsequently plant yield. Here we demonstrated that light fluctuation is part of the light signaling
489 pathway that regulates leaf senescence (Woo *et al.* 2019), and that stay green can be an informative
490 trait for the selection of varieties adapted to agroforestry systems. The reaction between light intensity
491 and reallocation patterns have already been observed in wheat with vegetative dry matter
492 redistribution into grains (Li *et al.* 2010), and with nitrogen reallocation in the leaves to optimize
493 photosynthetic performance exposure (Li *et al.* 2021).

494 Among others, it has been established that low light intensity under shading nets could affect the grain
495 quality traits, including the starch content its and composition (Li *et al.* 2006; Wang *et al.* 2013; Liu *et*
496 *al.* 2014) and grain appearance. Our results confirmed that light quantity affected starch content in
497 the grain, without altering the fraction of amylose. Wang et al. (2013) earlier reported some significant
498 differences in total starch content and amylose content of milled rice flour, and concluded to the
499 contribution of heredity, environment and their interaction for indica hybrid rice varieties under
500 shading. The decrease in the amount of starch and amylose was also claimed by Li et al. (2006) in two
501 grain varieties indica and japonica under shading treatments with nets (50% transmittance rate). The
502 presence of chalk, which is a key criterion for visual assessment of grain marketability, can also be
503 affected by shading environmental-related stress (Liu *et al.* 2014; Patindol *et al.* 2015). High porosity
504 is usually related to the presence of chalk (Singh *et al.* 2003), and characterized by the presence of
505 disorganized cellular structure (Lisle *et al.* 2000) leading to a greater cell and granular swelling,
506 detrimental to grain eating quality. The four studied varieties showed contrasted grain porosity in
507 responses to light quantity and fluctuation. The indica variety tended to present a lower porosity under
508 both shading treatments while the three other varieties showed either higher or lower porosity than
509 the control treatment. In this study, the significant loss in yield under shading treatments did not result
510 in a significant decrease in milling quality as it observed in previous studies (Liu *et al.* 2014). Further
511 research would be needed to investigate the qualitative response of rice plants to fluctuating
512 radiations and shading, whose tolerance might be genotype-dependent. It is likely to favor the
513 selection of suitable cultivars to meet the expectations of end-users, under constrained agroforestry
514 growing environment.

515 The photosynthetic capacity measured via light response curve highlighted that light quantity
516 and fluctuation triggered different responses among the four studied varieties. The high interaction
517 between varieties and shading treatments was already stressed in previous studies (Wang *et al.*
518 2015b; Pan *et al.* 2016), revealing the difficulty to select adapted varieties trough photosynthetic
519 measurements. The physiological measurements performed did not help retrieve the difference in

production among varieties, suggesting that yield is the consequence of multifactorial processes that can compensate variations in photosynthetic capacity. These specific responses to light intensity and fluctuation did not provide substantial benefits in term of production. Nevertheless, the absence of evolution of the morphological kernel traits under shading stressed that kernel shape is highly heritable and irresponsive to change in light intensity and fluctuation. It is likely that leaf scale measurements did not reflect an average value of the whole canopy photosynthesis. Besides, photosynthesis measurements are time consuming, preventing the efficient screening of varieties for breeding purpose.

Even if marked interactions were observed in physiological traits among varieties and treatments, the yield ranking between the four varieties was maintained in the three treatments. From an experimental and phenotyping perspective, this result tends to support that artificial light depletion using shade net could be sufficient to screen varieties of interest for agroforestry systems not yet implemented. Our result also confirmed previous findings revealing that productive materials in full sun conditions are also good producers in low radiation (Bueno & Lafarge 2017). The very specific behavior and low production (with potential low cooking and eating quality) of V7 together with the comparable production of the three other varieties made the identification of traits of interest difficult, without implementing reliable and discriminating descriptors. Indeed, the contrasted traits values between the two groups of individuals, formed by V7 and the three other varieties, could bias trait correlations. However, the variety V5 showed a significant higher production than the other varieties in the S-AF treatment, without exhibiting a chalky endosperm appearance potentially detrimental to its grain marketability. This variety maintained the highest harvest index among all the varieties and treatments, with a limited expansion of vegetative biomass in response to shade and a high leaf senescence under shade. These characteristics are consistent with the shade tolerant responses optimizing the trade-off between the elongation of vegetative parts and yield (Wille *et al.* 2017), and pave the way for selecting varieties dedicated to agroforestry systems. Further research would be needed to enlarge the diversity of the genotypes to screen and clearly identify if specific traits are better adapted to agroforestry systems. Other perspectives studies would also need to integrate other factors than the light resource, including a more global analysis with the environmental, social and economic aspects of the agroforestry system proposed (Rodenburg *et al.* 2022).

Acknowledgments

The authors would like to thank the Montpellier GAMET Biological Resource Centre (Centre de Ressource Centre de Ressources Biologiques des “Graines Adaptées aux conditions Méditerranéennes et Tropicales », ARCAD Montpellier – UMR AGAP Institute de Montpellier) for providing the

germplasm. The authors would like to thank Clémence Foray and Matthieu Dejean for the grain image acquisition, and for morphological script using open source ImageJ program, Romain Domingo for assistance in grain quality analyses. This research was supported by the General Directorate for Research & Strategy of CIRAD with incentive action “Scientific Creativity & Innovation” (CreSi) to both the Agroforice (RP), and the voxOmic projects (OG). RP and RV have been supported by the MaCS4Plants CIRAD network, initiated from the AGAP Institute and AMAP joint research units.

Reference

Acevedo-Siaca L.G., Coe R., Wang Y., Kromdijk J., Quick W.P. & Long S.P. (2020) Variation in photosynthetic induction between rice accessions and its potential for improving productivity. *New Phytologist*.

Ahirwal J., Sahoo U.K., Thangjam U. & Thong P. (2021) Oil palm agroforestry enhances crop yield and ecosystem carbon stock in northeast India: Implications for the United Nations Sustainable Development Goals. *Sustainable Production and Consumption*.

Alridiwirsah, Harahap E.M., Akoeb E.N. & Hanum H. (2019) Integrated cropping system of rice with oil palm: Local and new varieties. *Bulgarian Journal of Agricultural Science* **25**, 494–498.

Arenas-Corraliza M.G., Rolo V., López-Díaz M.L. & Moreno G. (2019) Wheat and barley can increase grain yield in shade through acclimation of physiological and morphological traits in Mediterranean conditions. *Scientific Reports* **9**, 1–10.

Ashraf M., Zulkifli R., Sanusi R., Tohiran K.A., Terhem R., Moslim R., … Azhar B. (2018) Alley-cropping system can boost arthropod biodiversity and ecosystem functions in oil palm plantations. *Agriculture, Ecosystems & Environment* **260**, 19–26.

Ashraf M., Sanusi R., Zulkifli R., Tohiran K.A., Moslim R., Ashton-Butt A. & Azhar B. (2019) Alley-cropping system increases vegetation heterogeneity and moderates extreme microclimates in oil palm plantations. *Agricultural and Forest Meteorology* **276–277**, 107632.

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

Bhagwat S.A. & Willis K.J. (2008) Agroforestry as a solution to the oil-palm debate. *Conservation Biology* **22**, 1368–1369.

Bueno C.S. & Lafarge T. (2017) Maturity groups and growing seasons as key sources of variation to consider within breeding programs for high yielding rice in the tropics. *Euphytica* **213**, 1–18.

Chen H., Li Q.P., Zeng Y.L., Deng F. & Ren W.J. (2019) Effect of different shading materials on grain yield and quality of rice. *Scientific Reports* **9**, 1–9.

Cheng F.M., Zhong L.J., Wang F. & Zhang G.P. (2005) Differences in cooking and eating properties between chalky and translucent parts in rice grains. *Food Chemistry* **90**, 39–46.

Dauzat J. & Eroy M.N. (1997) Simulating light regime and intercrop yields in coconut based farming systems. In *Developments in Crop Science*. pp. 87–98.

Dutta S.S., Tyagi W., Pale G., Pohlong J., Aochen C., Pandey A., … Rai M. (2018) Marker–trait association for low-light intensity tolerance in rice genotypes from Eastern India. *Molecular Genetics and Genomics* **293**, 1493–1506.

Franklin K.A. (2008) Shade avoidance. *New Phytologist* **179**, 930–944.

Gawankar M.S., Haldankar P.M., Salvi B.R. & Haldavanekar P.C. (2018) Intercropping in Young Oil Palm Plantation under Konkan Region of Maharastra , India. **7**, 2752–2761.

Gérard A., Wollni M., Hölscher D., Irawan B., Sundawati L., Teuscher M. & Kreft H. (2017) Oil-palm yields in diversified plantations: Initial results from a biodiversity enrichment experiment in Sumatra, Indonesia. *Agriculture, Ecosystems and Environment* **240**, 253–260.

Giraldo Toro A., Gibert O., Ricci J., Dufour D., Mestres C. & Bohuon P. (2015) Digestibility prediction of cooked plantain flour as a function of water content and temperature. *Carbohydrate Polymers* **118**, 257–265.

Haun J.R. (1973) Visual Quantification of Wheat Development 1. *Agronomy Journal* **65**, 116–119.

Heschel M.S., Stinchcombe J.R., Holsinger K.E. & Schmitt J. (2004) Natural selection on light response curve parameters in the herbaceous annual, Impatiens capensis. *Oecologia* **139**, 487–494.

Jia J., Xu M., Bei S., Zhang H., Xiao L., Gao Y., … Qiao X. (2021) Impact of reduced light intensity on wheat yield and quality: Implications for agroforestry systems. *Agroforestry Systems* **95**, 1689–1701.

Khasanah N., van Noordwijk M., Slingerland M., Sofiyudin M., Stomph D., Migeon A.F. & Hairiah K. (2020) Oil Palm Agroforestry Can Achieve Economic and Environmental Gains as Indicated by Multifunctional Land Equivalent Ratios. *Frontiers in Sustainable Food Systems* **3**, 1–13.

Koussihouèdé H., Clermont-Dauphin C., Aholoukpè H., Barthès B., Chapuis-Lardy L., Jassogne L. & Amadji G. (2020) Diversity and socio-economic aspects of oil palm agroforestry systems on the Allada plateau, southern Benin. *Agroforestry Systems* **94**, 41–56.

Lacube S., Fournier C., Palaffre C., Millet E.J., Tardieu F. & Parent B. (2017) Distinct controls of leaf widening and elongation by light and evaporative demand in maize. *Plant Cell and Environment* **40**, 2017–2028.

Lafarge T., Seassau C., Martin M., Bueno C., Clément-Vidal A., Schreck E. & Luquet D. (2010) Regulation and recovery of sink strength in rice plants grown under changes in light intensity. *Functional Plant Biology* **37**, 413.

Li H., Jiang D., Wollenweber B., Dai T. & Cao W. (2010) Effects of shading on morphology, physiology and grain yield of winter wheat. *European Journal of Agronomy* **33**, 267–275.

Li T., Ohsugi R., Yamagishi T. & Sasaki H. (2006) Effects of Weak Light on Starch Accumulation and Starch Synthesis Enzyme Activities in Rice at the Grain Filling Stage MATERIALS AND METHODS Materials Field experiments and indoor measurements were. **13**, 51–58.

Li Y.T., Yang C., Zhang Z.S., Zhao S.J. & Gao H.Y. (2021) Photosynthetic acclimation strategies in response to intermittent exposure to high light intensity in wheat (Triticum aestivum L.). *Environmental and Experimental Botany* **181**, 104275.

Lisle A.J., Martin M. & Fitzgerald M.A. (2000) Chalky and translucent rice grains differ in starch composition and structure and cooking properties. *Cereal Chemistry* **77**, 627–632.

Liu Q. hua, Wu X., Chen B. cong, Ma J. qing & Gao J. (2014) Effects of Low Light on Agronomic and Physiological Characteristics of Rice Including Grain Yield and Quality. *Rice Science* **21**, 243–251.

Masure A., Martin P., Lacan X. & Rafflegeau S. (2022) Promouvoir l'agroforesterie à base de palmiers à huile : un atout pour la durabilité de la filière. *Cahiers Agricultures* **31**, 14.

Murchie E.H., Hubbart S., Peng S. & Horton P. (2005) Acclimation of photosynthesis to high irradiance in rice: Gene expression and interactions with leaf development. *Journal of Experimental Botany* **56**, 449–460.

Naylor R.L., Battisti D.S., Vimont D.J., Falcon W.P. & Burke M.B. (2007) Assessing risks of climate variability and climate change for Indonesian rice agriculture. *Proceedings of the National Academy of Sciences* **104**, 7752–7757.

Pan S., Liu H., Mo Z., Patterson B., Duan M., Tian H., … Tang X. (2016) Effects of Nitrogen and Shading on Root Morphologies, Nutrient Accumulation, and Photosynthetic Parameters in Different Rice Genotypes. *Scientific Reports* **6**, 32148.

Patindol J.A., Siebenmorgen T.J. & Wang Y.J. (2015) Impact of environmental factors on rice starch structure: A review. *Starch/Staerke* **67**, 42–54.

Patindol J. & Wang Y.J. (2003) Fine structures and physicochemical properties of starches from chalky and translucent rice kernels. *Journal of Agricultural and Food Chemistry* **51**, 2777–2784.

Pérez E., Rolland-Sabaté A., Dufour D., Guzmán R., Tapia M., Raymundez M., … Gibert O. (2013) Isolated starches from yams (Dioscorea sp) grown at the Venezuelan Amazons: Structure and functional properties. *Carbohydrate Polymers* **98**, 650–658.

Perez R.P.A., Costes E., Théveny F., Griffon S., Caliman J.P. & Dauzat J. (2018) 3D plant model assessed by terrestrial LiDAR and hemispherical photographs: A useful tool for comparing light interception among oil palm progenies. *Agricultural and Forest Meteorology* **249**, 250–263.

Perez R.P.A., Pallas B., Le Moguédec G., Rey H., Griffon S., Caliman J.-P., … Dauzat J. (2016) Integrating mixed-effect models into an architectural plant model to simulate inter- and intra-progeny variability: a case study on oil palm (Elaeis guineensis Jacq.). *Journal of Experimental Botany* **67**, 4507–4521.

Perez R.P.A., Vezy R., Brancheriau L., Boudon F., Grand F., Ramel M., … Dauzat J. (2022) When architectural plasticity fails to counter the light competition imposed by planting design: an in silico approach using a functional–structural model of oil palm. *in silico Plants* **4**, 1–16.

Poorter H., Niinemets Ü., Ntagkas N., Siebenkäs A., Mäenpää M., Matsubara S. & Pons ThijsL. (2019) A meta-analysis of plant responses to light intensity for 70 traits ranging from molecules to whole plant performance. *New Phytologist* **223**, 1073–1105.

Qu M., Hamdani S., Li W., Wang S., Tang J., Chen Z., … Zhu X. (2016) Rapid stomatal response to fluctuating light: An under-explored mechanism to improve drought tolerance in rice. *Functional Plant Biology* **43**, 727–738.

Rodenburg J., Mollee E., Coe R. & Sinclair F. (2022) Global analysis of yield benefits and risks from integrating trees with rice and implications for agroforestry research in Africa. *Field Crops Research* **281**, 108504.

da Silva Maia R., Vasconcelos S.S., Viana-Junior A.B., Castellani D.C. & Kato O.R. (2021) Oil palm (Elaeis guineensis) shows higher mycorrhizal colonization when planted in agroforestry than in monoculture. *Agroforestry Systems* **95**, 731–740.

Singh N., Sodhi N.S., Kaur M. & Saxena S.K. (2003) Physico-chemical, morphological, thermal, cooking and textural properties of chalky and translucent rice kernels. *Food Chemistry* **82**, 433–439.

Susanti A., Marhaento H., Permadi D.B., Hermudananto, Budiadi, Imron M.A., … Lembasi M. (2020) Smallholder farmers' perception on oil palm agroforestry. *IOP Conference Series: Earth and Environmental Science* **449**.

Taniyoshi K., Tanaka Y. & Shiraiwa T. (2020) Genetic variation in the photosynthetic induction response in rice (Oryza sativa L.). *Plant Production Science* **23**, 513–521.

Tivet F., da Silveira Pinheiro B., de Rassac M. & Dingkuhn M. (2001) Leaf blade dimensions of rice (Oryza sativa L. and Oryza glaberrima Steud.). Relationships between tillers and the main stem. *Annals of Botany* **88**, 507–511.

Vialet-Chabrand S., Matthews J.S.A., Simkin A.J., Raines C.A. & Lawson T. (2017) Importance of fluctuations in light on plant photosynthetic acclimation. *Plant Physiology* **173**, 2163–2179.

Wang L., Deng F. & Ren W.J. (2015a) Shading tolerance in rice is related to better light harvesting and use efficiency and grain filling rate during grain filling period. *Field Crops Research* **180**, 54–62.

Wang L., Deng F. & Ren W.J. (2015b) Shading tolerance in rice is related to better light harvesting and use efficiency and grain filling rate during grain filling period. *Field Crops Research* **180**, 54–62.

Wang L., Deng F., Ren W.J. & Yang W.Y. (2013) Effects of Shading on Starch Pasting Characteristics of Indica Hybrid Rice (Oryza sativa L.). *PLoS ONE* **8**.

Wangpakapattanawong P., Finlayson R. & Öborn I. (2017) *Agroforestry in rice-production landscapes in Southeast Asia a practical manual*. FAO Regional Office for Asia and the Pacific.

Wille W., Pipper C.B., Rosenqvist E., Andersen S.B. & Weiner J. (2017) Reducing shade avoidance responses in a cereal crop. *AoB PLANTS* **9**, 1–16.

Woo H.R., Kim H.J., Lim P.O. & Nam H.G. (2019) Leaf Senescence: Systems and Dynamics Aspects. *Annual Review of Plant Biology* **70**, 347–376.

Yang F., Feng L., Liu Q., Wu X., Fan Y., Raza M.A., … Yang W. (2018) Effect of interactions between light intensity and red-to-far-red ratio on the photosynthesis of soybean leaves under shade condition. *Environmental and Experimental Botany* **150**, 79–87.

Zeileis A. & Hothorn T. (2002) Diagnostic Checking in Regression Relationships. *R News* **2**, 7--10.

Zemp D.C., Ehbrecht M., Seidel D., Ammer C., Craven D., Erkelenz J., … Kreft H. (2019) Mixed-species tree plantings enhance structural complexity in oil palm plantations. *Agriculture, Ecosystems and Environment* **283**, 106564.

Zhang N., van Westreenen A., Anten N.P.R., Evers J.B. & Marcelis L.F.M. (2020) Disentangling the effects of photosynthetically active radiation and red to far-red ratio on plant photosynthesis under canopy shading: a simulation study using a functional-structural plant model. *Annals of botany* **126**, 635–646.

Zhen S. & Bugbee B. (2020) Far-red photons have equivalent efficiency to traditional photosynthetic photons: Implications for redefining photosynthetically active radiation. *Plant Cell and Environment* **43**, 1259–1272.