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# Modeling functional relationships between morphogenetically active radiation and photosynthetic photon flux density in mango tree crown

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Light is a key factor in plant ecophysiological modeling because of its crucial effects on plant growth and development. However, solar light quantity and quality change with environmental factors such as sky condition and solar elevation. When passing through a tree crown, light is modified by its interaction with the phytoelements, leaves and axes. This leads to a variability of light quantity and quality within the crown, with consequences on light-related processes such as photosynthesis and photomorphogenesis. We evaluated the effects of positional (depth within the crown) and environmental (sky condition, solar elevation) factors on light quantity and quality within the crown of the tropical evergreen mango tree. Functional relationships were modeled between morphogenetically active radiation variables that describe light quality [narrowband red ( $R_n$ ), narrowband far-red ( $FR_n$ ), the ratio  $\zeta = R_n : FR_n$ , and UVA-blue ( $UVA-BL$ )] and light quantity [photosynthetic photon flux density ( $PPFD$ ) and relative transmitted  $PPFD$  ( $TrPPFD$ )]. Light quantity and quality varied within the mango tree crown in a wide range similar to that of a forest. This variability was structured by the depth within the crown as well as by sky condition and solar elevation. Linear relationships linked  $R_n$ ,  $FR_n$  and  $UVA-BL$  to  $PPFD$ , and non-linear relationships linked  $\zeta$  to  $TrPPFD$ . These relationships were strong, accurate and unbiased. They were affected by positional and environmental factors. The results suggested that these relationships were shaped by the characteristics of incident solar light and/or by the interactions between light and phytoelements. Two consequences of interest emerged from this research: i) the modeled relationships allow to infer light quality, that is difficult and time-consuming to simulate, from light quantity modeling within a tree crown, and ii) sky condition and solar elevation should be considered to improve light modeling within a tree crown.

## KEYWORDS

crown depth, light spectrum, *Mangifera indica*, modeling, red : far-red ratio, sky condition, solar elevation

## 1 Introduction

Light is a key factor for plant growth and development, and is therefore of major importance for ecophysiological modeling (DeJong et al., 2011). It is characterized by two dimensions, light quantity, i.e., the quantity of photons per area and time units, that is relevant for photosynthesis, and light quality, i.e., the spectral composition of light, that is relevant for photomorphogenesis and photosynthesis. Modeling ecophysiological and developmental light-related processes in plant models requires an accurate estimation of light quantity and quality at the level of specific organs (e.g., leaves, buds), which might be difficult in complex architecture like trees. The difficulty arises from two points. First, the quantity and quality of incident solar light may vary with environmental conditions, and light quantity and quality vary within the tree (Durand et al., 2021). Second, light models focus mostly on simulating the quantity of light at the level of plant organs. Simulating light quality is more complex and time-consuming (Henke and Buck-Sorlin, 2017) since the whole spectral composition of each ray has to be computed with wavelength-specific processes during the interactions with each organ. Consequently, characterizing the effects of factors affecting light quantity and quality from the sun to the organs, and modeling light quality from light quantity would help improving the simulation of light for ecophysiological and developmental plant models.

Solar radiation interacts with tree crown phytoelements, i.e., the leaves and the stems, by wavelength-selective reflection, transmission and absorption of photons. The proportion of light components (direct and diffuse beams from above the canopy, scatterings from within the canopy) is affected by the crown structure (i.e., the geometry and arrangement of the phytoelements) and by the optical properties of the phytoelements (Grant, 1997; Kim et al., 2011). The optical properties of phytoelements are partly related to their species-specific composition and concentration in pigments, including chlorophyll. In this way, leaves absorb radiation in blue and red wavelengths and reflect or transmit most far-red wavelengths (Casal, 2013), leading to a simultaneous change in the quantity and quality of light at any point within and below a crown or canopy. Light modification mostly occurs in the range of photosynthetically active radiation (*PAR*, 400–700 nm) and of morphogenetically active radiation (*MAR*, 350–800 nm) (Varlet-Grancher et al., 1993; Combes et al., 2000). The simultaneous modification of light quantity and quality can be modeled by functional relationships between variables related to light quality (i.e., the photon flux density integrated into bands specific to the *MAR* variables or the ratio between red and far-red photon flux density (see below)) and variables related to light quantity (the photosynthetic photon flux density (*PPFD*), i.e., the photon flux density integrated into the *PAR* band, or the relative transmitted *PPFD* (*TrPPFD*), i.e., the ratio between *PPFD* within the vegetation and incident *PPFD*). These relationships have been mostly described in temperate forests (Muraoka et al., 2001; Pecot et al., 2005; Leuchner et al., 2007), in a wet tropical forest (Capers and Chazdon, 2004), and more rarely in a stand of a single annual species (Escobar-Gutiérrez et al., 2009) or within a single tree crown (Combes and Escobar-Gutiérrez, 2009). The functional

relationships can vary with species and crown characteristics, e.g., between deciduous and coniferous stands (Leuchner et al., 2007).

*MAR* is perceived by different plant photoreceptors, mainly phytochromes and cryptochromes (Casal, 2013). Phytochromes are sensitive to the narrowband red (*Rn*: 655–665 nm) and the narrowband far-red (*FRn*: 725–735 nm) wavelengths. The *Rn* : *FRn* ratio, denoted  $\zeta$ , regulates various adaptive responses such as seed germination, stem growth, leaf expansion, leaf senescence, and branching (Franklin, 2008; Demotes-Mainard et al., 2016). Cryptochromes are sensitive to the *UVA-BL* wavelengths (350–500 nm). *UVA-BL* promotes bud burst, stomatal opening and regulates the time of flowering (Leduc et al., 2014; Brelsford and Robson, 2018). Furthermore, far-red wavelengths could achieve more efficient photosynthesis under fluctuating light (Kono et al., 2020) and could enhance photosynthesis when combined with shorter wavelengths (Zhen et al., 2021). Thus, the knowledge of light distribution (i.e., range of values taken by variables that describe light quantity and quality) within a plant or a stand is of great importance when studying or modeling light-related processes such as photosynthesis and photomorphogenesis. Recent studies aiming at improving photosynthesis efficiency under fluctuating light environment (Slattery et al., 2018; De Souza et al., 2022; Long et al., 2022) could benefit from such detailed light models. Most of the studies that focus on light distribution are related to ecology and describe *PPFD* and *MAR* under deciduous, coniferous and mixed temperate forests, as well as under tropical forests (Chazdon and Fetcher, 1984; Turnbull and Yates, 1993; Capers and Chazdon, 2004; Navrátil et al., 2007; Hertel et al., 2011). Studies measuring light distribution in annual crops (Chelle et al., 2007) or individual deciduous (Combes and Escobar-Gutiérrez, 2009) or evergreen (Ladux et al., 2021) tree crowns are rare. Within an evergreen tree crown, the contrasting light environment between the outer and the inner part of the crown is permanent throughout the year, with probable effects on bud burst and architectural development, contrary to deciduous trees where light environment, at the time of bud burst in spring, varies little between the buds which are exposed to direct solar light.

Light within the plant stand or tree crown depends on the characteristics of incident solar light, which is influenced by environmental factors such as solar elevation and atmospheric composition. The quantity of incident light above the canopy follows the solar elevation angle  $\alpha$  during the day. Light quality may also vary throughout the day. At low solar elevation ( $\alpha < 20^\circ$ ), the fraction of blue wavelengths is higher than at high solar elevation ( $\alpha > 20^\circ$ ), whereas the opposite is observed for the fraction of red wavelengths (Chiang et al., 2019). Similarly, high variability of  $\zeta$  values is observed at twilight, compared to stable values during the rest of the day (Kotilainen et al., 2020). The effect of solar elevation can vary with the position within the canopy and the atmospheric composition (Navrátil et al., 2007). Atmosphere with more clouds and/or higher aerosol concentration has a higher proportion of diffuse light because of scatterings. In this way, clouds reduce light quantity compared to clear sky and modify light quality. Under cloudy sky, as compared to clear sky, incident light is depleted in the red and far-red wavelengths and is enriched in the UV and blue wavelengths (Durand et al., 2021).

The ratio  $\zeta$  of the incident light remains constant or slightly increases under cloudy sky (Chiang et al., 2019). However, within the tree canopy,  $\zeta$  has significantly higher values under cloudy sky compared to clear sky (Leuchner et al., 2007; Hertel et al., 2011; Dengel et al., 2015). Thus, since solar elevation and sky conditions (cloudiness, aerosols) affect the characteristics of incident solar light, they may alter the relationships between light quality and quantity within the plant stand (Capers and Chazdon, 2004; Pecot et al., 2005; Leuchner et al., 2007) and should therefore be considered when modeling these relationships.

Mango tree (*Mangifera indica* L.) is a tropical evergreen tree of economic, nutritional and cultural importance in many countries throughout the tropics. Its crown is roughly spherical and formed by a dense, about 1-m-thick, layer of leafy axes (Figure 1A). Leafless scaffolds and branches are in the central part of the crown (Figure 1B). To quantify light distribution and model the functional relationships between *MAR* and *PAR* variables within this evergreen fruit tree crown would provide a better understanding of the range and variability of light environment within a complex tree crown. Furthermore, it would improve the simulation of light quantity and quality in the functional-structural plant model V-Mango (Boudon et al., 2020; Vaillant et al., 2022), contributing to improve the modeling of photosynthesis and photomorphogenic processes. We addressed three main questions in this study based on an extensive data set collected in the field. What are the ranges of light quantity and quality within the mango tree crown under different environmental conditions (solar elevation and sky condition)? How can be modeled the functional relationships linking light quality and quantity? How are these relationships affected by environmental conditions and position within the crown?

## 2 Materials and methods

### 2.1 Study site and plant material

The study site is an experimental orchard located at the CIRAD (French Agricultural Research Center for International Development) research station at Saint-Pierre, Reunion Island (21°06'S, 55°32'E,

125 m a.s.l.). The experimental orchard is composed of 153 mango trees (cultivar 'Cogshall' grafted onto 'Maison Rouge' rootstock). Trees were planted in 2004 with a spacing of 5 m between trees along the row and 6 m between rows. Row orientation was northeast-southwest (34° from the north). Trees were pruned every two years starting in 2012. Irrigation was applied with a drip system on an evapotranspiration basis (about 5 mm d<sup>-1</sup> on average) as of flowering, until rainfall was sufficient to maintain soil moisture.

Light measurements were performed during the hot and rainy season, from December 18, 2019, to February 17, 2020, on four mango trees randomly selected in the orchard. The four trees were (mean ± s.d.) 4.5 ± 0.4 m high, and their canopy diameter was 5.0 ± 0.3 m along the row and 5.3 ± 0.2 m between rows.

### 2.2 Experimental design

In order to capture the variability of light quantity and quality within the mango tree crown and the effects of environmental conditions, light was measured on a sample of terminal and non-terminal growth units with different positions within the crown, and under different solar elevations and sky conditions. A growth unit (GU) is a leafy portion of axis stemming from a vegetative bud and whose extension is uninterrupted (Figure 1C). In the mango tree, GU extension lasts 8 to 12 days (Dambreville et al., 2013). Twelve GUs were sampled per tree at three different positions (hereafter called 'depth') defined according to their depth within the leafy layer of the crown (outer part of the crown, middle leafy part of the crown, inner part of the crown without leaves). For each depth modality, four GUs were sampled at each of four orientation zones (the four zones are defined as the two opposite crown sides of two orientations: on the diameter along the row and on the diameter perpendicular to the row).

The environmental conditions consisted of three solar elevations and two sky conditions. Solar elevation was referred to as the local time of measurement (7:00, 12:00 and 17:00, UTC +4:00). Average solar elevations during the measurement campaign were 22.98° at 7:00, 85.16° at 12:00 and 19.58° at 17:00. Sky conditions (clear and cloudy) were defined according to the average percentage of diffused light in incident light during a



FIGURE 1

Photographs illustrating adult mango tree, cultivar Cogshall, presenting an exterior spherical outer crown (A) and a leafless shaded inner crown (B) (hemispherical photograph taken at dawn, positioned above the trunk at 1.20 m high). Measurements were done on the apical end of sampled growth units (C) at each depth within the crown (outer, middle, and inner parts) and orientation (along the row and perpendicular to the row).



measurement series. A series consisted of the 48 light measurements made the same day at a given solar elevation on the 12 GUs of the four trees. Measurement series with an average percentage of diffuse light greater (resp. lower) than 40% were classified as a cloudy (resp. clear) sky condition. The relevance of this sky classification based on the percentage of diffuse light was checked with a second indicator, the clearness index  $Kt$ . It is a dimensionless number ranging from 0 to 1, with low (resp. high) values under cloudy (resp. clear) sky conditions. Methods to measure and calculate the two indicators are described in [Supplementary Material 1](#). Across our measurement series, the average percentage of diffused light was 21% under clear skies and 63% under cloudy skies, with values ranging from 9%–40% and 45%–77%, respectively. Measurement series were associated with an average  $Kt = 0.67$  under clear sky and an average  $Kt = 0.30$  under cloudy sky, with values ranging from 0.56–0.81 and 0.17–0.48, respectively. The sky classification of the measurement series therefore appeared to be relevant on the basis of the  $Kt$  values since  $Kt$  was  $> 0.5$  (resp.  $< 0.5$ ) for all series under clear (resp. cloudy) sky conditions ([Supplementary Figure S1](#)).

A total of 31 light measurement series, corresponding to 1488 light spectra, were acquired during the campaign. Series were split into two independent datasets. The first dataset (Dataset 1) consisted of 18 measurement series (three series per solar elevation  $\times$  sky condition combination) with a total of 864 light spectra. The second dataset (Dataset 2) consisted of the remaining 13 measurement series, with a total of 624 light spectra. Dataset 1 was a balanced combination of the depth, solar elevation and sky condition factor levels (3 series  $\times$  4 trees  $\times$  4 GUs  $\times$  3 depth  $\times$  3 solar elevations  $\times$  2 sky conditions) and was used for model calibration. On the contrary, Dataset 2 was unbalanced for solar elevation and sky condition factors (there were four series at 7:00 under clear sky, five series at 12:00 under clear sky, two series at 17:00 under clear sky and two series at 17:00 under cloudy sky) and was used for model validation.

## 2.3 Light measurement and calculation of PAR and MAR variables

Light spectral irradiance ( $\mu\text{W cm}^{-2} \text{ nm}^{-1}$ ) was measured with a JAZ spectrometer (Ocean Insight, USA), with wavelength ranging from 200 to 1100 nm and a spectral resolution of 1 nm. The spectrometer was connected to an optical fiber and a cosine corrector (i.e., optical diffuser) to collect signal with a 180° field of view. Spectrometer was adapted to solar light and to dark using calibration files before measurements. After calibration, the sensor was placed above the apical end of the sampled GUs and maintained horizontally during the measurement thanks to a gimbal stabilizer. At the same time, the incident  $PPFD$  ( $\mu\text{mol m}^{-2} \text{ s}^{-1}$ ) received horizontally above the canopy was measured in an open field at about 100 m from the experimental orchard using a PQS1 PAR quantum sensor (Kipp & Zonen, the Netherlands) connected to a CR1000 X (Campbell, USA) data logger. The output signal of the PQS1 PAR quantum sensor was measured at 5-s intervals and averaged every 30 s. Before the series

of measurements, a relationship was calibrated between the incident  $PPFD$  measured with the JAZ spectrometer and measured with the PQS1 PAR quantum sensor for each sky condition and at different hours between 7:00 and 18:00 to cover the full range of incident  $PPFD$  ([Supplementary Material 1](#); [Supplementary Figure S2](#)). This relationship was used to correct the measurement obtained with the PQS1 PAR quantum sensor on the JAZ spectrometer for each series.

Based on the spectral irradiance measured in energy ( $\mu\text{W cm}^{-2} \text{ nm}^{-1}$ ) with the JAZ spectrometer, irradiance in photon flux density ( $\mu\text{mol m}^{-2} \text{ s}^{-1}$ ) was calculated for each wavelength using series of conversions and considering the geographical location of the site and the integration time recorded in JAZ files, according to the manufacturer's instructions described in [Supplementary Material 2](#). Then,  $PPFD$ ,  $UVA-BL$ ,  $Rn$  and  $FRn$  variables ( $\mu\text{mol m}^{-2} \text{ s}^{-1}$ ) were calculated by integrating the photon flux density in the PAR band (400–700 nm;  $PPFD$ ), UVA-blue band (350–500 nm;  $UVA-BL$ ), narrowband red (655–665 nm;  $Rn$ ) and narrowband far-red (725–735 nm;  $FRn$ ). The variable  $\zeta$  (dimensionless) was calculated as the ratio between  $Rn$  and  $FRn$ . Relative transmitted  $PPFD$  ( $TrPPFD$ ) was calculated as the ratio between the  $PPFD$  measured with the spectrometer in the mango tree crown and the incident  $PPFD$  measured at the same time with the PQS1 PAR quantum sensor.

## 2.4 Data analysis

### 2.4.1 Light variability in the mango tree crown in response to positional and environmental factors

Three-way analyses of variance (ANOVAs) were used to test the effects of the positional (depth within the crown) and environmental (sky condition and solar elevation) factors and their second- and third-order interactions on the six light variables ( $PPFD$ ,  $TrPPFD$ ,  $UVA-BL$ ,  $Rn$ ,  $FRn$  and  $\zeta$ ). Light variables were log-transformed prior to analysis to conform to normality, except  $\zeta$  for which a square-root-transformation was applied. Analyses were performed on Dataset 1.

### 2.4.2 Functional relationships between MAR and PAR variables

#### 2.4.2.1 Calibration and validation of the functional relationships

$UVA-BL$ ,  $Rn$  and  $FRn$  were related to  $PPFD$  with linear models (Equation 1).

$$y = a + b \text{ } PPFD \quad (\text{Equation 1})$$

where  $y$  represents  $UVA-BL$ ,  $Rn$  or  $FRn$ ,  $a$  ( $\mu\text{mol m}^{-2} \text{ s}^{-1}$ ) is the y-intercept, and  $b$  (dimensionless) is the slope. The y-intercept  $a$  was set at 0 for  $Rn$  since the wavelength range of this variable is totally included in the PAR band; consequently,  $y = 0$  when  $PPFD = 0$ . On the contrary, the wavelength range of  $UVA-BL$  and  $FRn$  is partially or completely, respectively, outside the PAR band. In such a case,  $y$  might be  $> 0$  when  $PPFD = 0$  due to the contribution of broadband irradiance. Therefore, the y-intercept  $a$  was not set at 0 for  $UVA-BL$  and  $FRn$ .

The relationship between  $\zeta$  and  $TrPPFD$  was non-linear and modeled using the Schnute equation ([Schnute, 1981](#)) (Equation 2).

$$\zeta = \left[ c^b + (d^b - c^b) \frac{1 - e^{-a(TrPPFD-x_1)}}{1 - e^{-a(x_2-x_1)}} \right]^{1/b} \tag{Equation 2}$$

where parameters  $a$  and  $b$  define the shape of the curve,  $x_1$  and  $x_2$  are the first and last values of the  $TrPPFD$  domain for which the model was fitted, and parameters  $c$  and  $d$  are the values of  $\zeta$  at  $x_1$  and  $x_2$ , respectively. All parameters are dimensionless. On the basis of the range of  $TrPPFD$  measurements (see results), the lower and upper bounds for  $TrPPFD$ ,  $x_1$  and  $x_2$ , were set to 0 and 1, respectively. When  $TrPPFD = x_1 = 0$  (i.e., no light between 400 and 700 nm),  $Rn = 0$  (since  $Rn$  range is included in  $PPFD$  range) and  $FRn > 0$  (since  $FRn$  range is outside the  $PPFD$  range) then  $\zeta = 0$ . The parameter  $c$  was therefore set to 0. Consequently, the Schnute equation was simplified as follows:

$$\zeta = d \left[ \frac{1 - e^{-a TrPPFD}}{1 - e^{-a}} \right]^{1/b} \tag{Equation 3}$$

The functional relationships were calibrated globally, i.e., independently of the positional and environmental factors, on Dataset 1 using the least-squares method.

Model goodness-of-fit was assessed on the calibration data (Dataset 1) with calculation of model bias (systematic errors) and accuracy (random errors). A common bias measurement is the average error  $ME$  (Equation 4; Walther and Moore, 2005). It indicates if the model, on average, overestimates ( $ME > 0$ ) or underestimates ( $ME < 0$ ) the data. The expected value is 0. A scaled statistic was used to assess model accuracy in terms of the percentage relative root mean square error with mean normalization ( $RRMSE$ ) (Equation 5). Model accuracy is considered excellent when  $RRMSE < 10\%$ , good if  $10\% < RRMSE < 20\%$ , fair if  $20\% < RRMSE < 30\%$ , and poor if  $RRMSE > 30\%$  (Despotovic et al., 2016).

$$ME = \frac{1}{n} \sum_{i=1}^n (\hat{y}_i - y_i) \tag{Equation 4}$$

$$RRMSE = \frac{100}{\bar{y}} \sqrt{\frac{1}{n} \sum_{i=1}^n (\hat{y}_i - y_i)^2} \tag{Equation 5}$$

where  $y_i$  is the  $i^{th}$  observed value,  $\hat{y}_i$  is the corresponding value estimated by the model,  $\bar{y}$  is the mean of observed values, and  $n$  is the number of observed data.

The models were also evaluated by assessing their predictive quality on independent data (Dataset 2). Values predicted with the model were compared to observed values using the previous statistics ( $ME$  and  $RRMSE$ ) and a plot analysis. The regression line between predicted and observed values was computed. The expected values for the slope, the y-intercept and the coefficient of determination of the regression are 1, 0 and 1, respectively. It was then checked if 1 and 0 were included in the 95% confidence interval of the slope and y-intercept, respectively.

### 2.4.2.2 Effect of positional and environmental factors on the functional relationships

The effects of positional (depth within the crown) and environmental (sky condition and solar elevation) factors on the linear relationships linking  $UVA-BL$ ,  $Rn$  and  $FRn$  to  $PPFD$

(Equation 1) were tested using analyses of covariance (ANCOVAs), with the studied  $MAR$  variable as the response variable,  $PPFD$  as the independent variable, and the tested factor as the covariable. The interaction between  $PPFD$  and the factor was tested first. A significant interaction indicated an effect of the factor on the slope, and no conclusion could be drawn on the effect of the factor on the y-intercept in this case. If the interaction was not significant, then the analysis of covariance was performed without interaction to test for the effect of the factor on the y-intercept. When a significant effect of a factor with more than two levels was detected on the slope (or y-intercept) of the relationship, pairwise comparisons were performed with Bonferroni P-value adjustment for multiple comparisons in order to identify groups of homogeneous slopes (or y-intercepts).

The effects of positional and environmental factors on the non-linear relationship between  $\zeta$  and  $TrPPFD$ , modeled with the simplified Schnute equation (Equation 3), were analyzed using the likelihood ratio methodology described in Kimura (1980). Kimura's likelihood ratio test (LRT) provides a general method for the statistical comparison of non-linear curves fitted to data, under the assumption of independent and normally distributed errors. In our case, considering a factor with  $I$  levels, each following the Schnute equation with parameters  $(a_i, b_i, d_i)$ ,  $i \in \{1, \dots, I\}$ , a base model ( $\Omega$ ) was specified as follows:  $\zeta_{i,j} = f_S(TrPPFD_{i,j}, a_i, b_i, d_i)$ , where  $f_S$  referred to the Schnute equation and  $j$  is the  $j^{th}$  observation in the  $i^{th}$  factor level. Assuming that all curves were coincident, i.e., the parameters  $(a_i, b_i, d_i)$  were not significantly different for the  $I$  factor levels, a coincident model ( $\omega$ ) was specified by adding the following set of  $r = 3(I - 1)$  equality constraints on parameters of base model  $\Omega$ :

$$\begin{cases} a_i = a_k \\ b_i = b_k, \quad \forall i \in \{1, \dots, I\}, \forall k \in \{1, \dots, I\} \cap k \neq i. \\ d_i = d_k \end{cases}$$

Testing the hypothesis that the factor had no effect on the relationship between  $\zeta$  and  $TrPPFD$  was thus equivalent to testing the hypothesis that all curves were coincident. The test is based on the LRT statistic, which approximately follows a  $\chi^2$  distribution:

$$LRT = -N \log \left( \frac{RSS_{\Omega}}{RSS_{\omega}} \right) \sim \chi_{df}^2 \tag{Equation 6}$$

where  $df$  is the number of degrees of freedom (equal to the number  $r$  of linear constraints),  $N$  is the total number of observations (all levels combined),  $RSS_{\Omega}$  is the total residual sum of squares derived from fitting the base model  $\Omega$  (curves were fitted separately for each factor level and their residual sums of squares were summed), and  $RSS_{\omega}$  is the residual sum of squares derived from fitting the coincident model  $\omega$  with the hypothesized constraints. When a significant effect of a factor with more than two levels was detected, the same LRT approach was used to perform pairwise comparisons with Bonferroni P-value adjustment for multiple comparisons in order to identify which pair(s) of curves significantly differed.

Covariance and LRT analyses described above were performed on Dataset 1 in two steps. First, the analyses were performed

separately for each factor. All data were used when testing the effects of sky condition and solar elevation. However, only the data subset corresponding to the highest solar elevation (local time = 12:00) was used when testing the effect of depth within the crown in order to have direct, vertical and unmodified light beams reaching the tree crown, and without light beams passing through the crown of the adjacent tree crowns at lower solar elevation. Second, the effect of solar elevation was separately tested under clear and cloudy sky conditions because it was assumed that the effect of solar elevation on the functional relationships was more likely under the clear than the cloudy sky condition. Indeed, directional distribution of light beams is anisotropic under the clear sky condition and isotropic under the cloudy sky condition. Light beams are therefore more homogeneous between the different solar elevations under cloudy than clear sky.

Statistical analyses were performed with R software, version 4.1.2 (R Core Team, 2021). Pairwise comparisons of estimated slopes and y-intercepts were performed with the *emtrends()* and *emmeans()* functions, respectively, of the *emmeans* package (Lenth, 2016). The Schnute non-linear equation was fitted with the *nxmlb()* function in the *nlmrt* package (Nash, 2014). Because of the large sample size and to highlight the main factors affecting light variables and functional relationships, the significance level in statistical tests was set at  $P = 0.01$ .

### 3 Results

#### 3.1 Light variability in the mango tree crown in response to positional and environmental factors

During the measurement series (Dataset 1), incident *PPFD* ranged from 119 to 2383  $\mu\text{mol m}^{-2} \text{s}^{-1}$ , with an average ( $\pm$  standard error, s.e.) value of 934 ( $\pm 17$ )  $\mu\text{mol m}^{-2} \text{s}^{-1}$ . *PPFD* measured within the mango tree crown ranged from 1 to 2221  $\mu\text{mol m}^{-2} \text{s}^{-1}$ , with an average value of 136 ( $\pm 10$ )  $\mu\text{mol m}^{-2} \text{s}^{-1}$ . *TrPPFD*

ranged from 0.001 to 0.94, with an average value of 0.107 ( $\pm 0.005$ ). *UVA-BL* ranged from 0.2 to 767.5  $\mu\text{mol m}^{-2} \text{s}^{-1}$ , with an average value of 47.8 ( $\pm 3.7$ )  $\mu\text{mol m}^{-2} \text{s}^{-1}$ . *Rn* ranged from 0.01 to 77.6  $\mu\text{mol m}^{-2} \text{s}^{-1}$  with an average value of 4.66 ( $\pm 0.36$ )  $\mu\text{mol m}^{-2} \text{s}^{-1}$ . *FRn* ranged from 0.10 to 73.0  $\mu\text{mol m}^{-2} \text{s}^{-1}$  with an average value of 5.02 ( $\pm 0.32$ )  $\mu\text{mol m}^{-2} \text{s}^{-1}$ . Values of  $\zeta$  varied from 0.02 to 1.27, with an average value of 0.53  $\pm 0.01$ .

Depth within the crown, sky condition and solar elevation significantly affected the *PAR* and *MAR* variables within the mango tree crown (Table 1; Figure 2). The three-way interaction ‘sky condition : solar elevation : depth within the crown’ and the two-way interaction ‘sky condition : depth within the crown’ were not significant for any variable. On the contrary, the depth within the crown had a significant and marked effect on the six studied variables (Table 1).

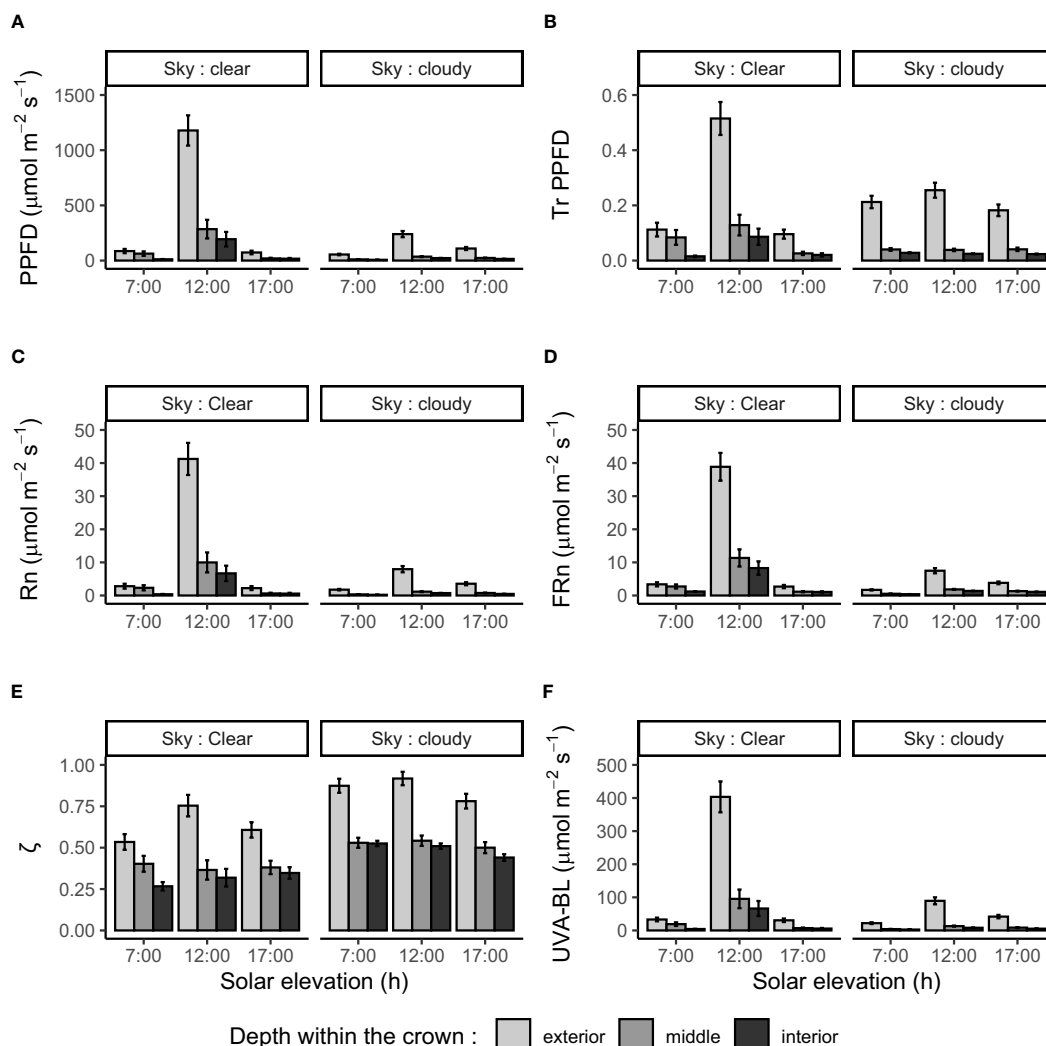
*PPFD*, *Rn*, *FRn* and *UVA-BL* displayed a very similar pattern according to the studied factors (Figures 2A, C, D, F). Those variables were markedly affected by depth within the crown and solar elevation (Table 1). As a general rule, average values were higher in the outer part of the tree crown, in particular, at high solar elevation (12:00). Two interactions were significant for these variables. The first one, ‘sky condition : solar elevation’, expressed the fact that average values for each solar elevation were less variable under cloudy sky than under clear sky where the average value at high solar elevation (12:00) was much higher than those at low solar elevation (7:00 and 17:00). The second interaction, ‘solar elevation : depth within the crown’ (which, however, was barely significant for *UVA-BL*; Table 1), expressed the fact that average values varied more with solar elevation in the outer part of the crown than within the crown (middle and inner parts). In particular, average values were higher at high solar elevation (12:00).

*TrPPFD* was markedly affected by depth within the crown (*TrPPFD* was on average higher in the outer part of the tree crown), by sky condition (*TrPPFD* was on average higher under clear sky), and by solar elevation (*TrPPFD* was on average higher at high solar elevation). The ‘solar elevation : depth within the crown’ interaction was significant. Overall, *PPFD*, *Rn*, *FRn*, *UVA-BL* and

TABLE 1 *F*- and *P*-values of the three-way ANOVAs testing the effects of sky condition (Sky.), solar elevation (Elev.) and depth within the crown (Depth) on photosynthetic photon flux density (*PPFD*), relative transmitted *PPFD* (*TrPPFD*), narrowband red (*Rn*), narrowband far-red (*FRn*), *Rn* : *FRn* ratio ( $\zeta$ ), and UVA-blue (*UVA-BL*) measured within four mango tree crowns.

Factors	df	<i>PPFD</i>		<i>TrPPFD</i>		<i>Rn</i>		<i>FRn</i>		$\zeta$		<i>UVA-BL</i>	
		<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>
Sky.	1	6.3	0.012	38.7	<0.001	0.5	0.462	136.8	<0.001	134.6	<0.001	13.9	<0.001
Elev.	2	105.4	<0.001	6.2	0.002	95.6	<0.001	276.0	<0.001	0.6	0.543	94.2	<0.001
Depth	2	157.5	<0.001	166.1	<0.001	136.3	<0.001	151.7	<0.001	103.0	<0.001	173.5	<0.001
Sky.: Elev.	2	15.7	<0.001	2.1	0.121	14.8	<0.001	67.6	<0.001	3.6	0.026	14.9	<0.001
Sky.: Depth	2	0.4	0.665	0.4	0.640	0.4	0.683	2.1	0.125	0.6	0.547	0.4	0.656
Elev.: Depth	4	3.5	0.008	3.7	0.006	4.0	0.003	4.8	<0.001	2.0	0.090	2.8	0.025
Sky.: Elev.: Depth	4	1.4	0.223	1.5	0.207	1.8	0.117	1.7	0.140	1.3	0.266	1.1	0.337
Residuals	846												

Variables were log-transformed prior to analysis, except  $\zeta$  for which a square-root transformation was applied. *P*-values in bold are significant at  $P < 0.01$ .



**FIGURE 2**  
 Photosynthetic photon flux density (*PPFD*; **A**), relative transmitted *PPFD* (*TrPPFD*; **B**), narrowband red (*Rn*; **C**), narrowband far-red (*FRn*; **D**), *Rn* : *FRn* ratio ( $\zeta$ ; **E**), and UVA-blue (*UVA-BL*; **F**) measured within four mango tree crowns according to sky condition (clear or cloudy), solar elevation (referred to as local time: 7:00, 12:00, 17:00) and depth within the crown (outer, middle and inner parts). Bars and segments represent means and standard errors ( $n = 48$ ) of observed data, respectively.

*TrPPFD* varied more with solar elevation in the outer part of the crown (with the higher values at 12:00) and under clear sky than under cloudy sky.

The variable  $\zeta$  was affected by sky condition ( $\zeta$  was on average higher under cloudy sky) and by depth within the crown ( $\zeta$  was on average higher at the outer part of the crown). Solar elevation had no significant effect on  $\zeta$ , and no interaction was significant.

### 3.2 Functional relationships between MAR and PAR variables

#### 3.2.1 Calibration and validation of the functional relationships

Fits of the linear models linking *MAR* variables *UVA-BL*, *Rn* and *FRn* to *PPFD* on the calibration Dataset 1 were highly significant ( $r^2 > 0.99$ ;  $P < 0.001$ ), with little variability around the

fit (Table 2; Supplementary Figure S3). Goodness-of-fit indicators showed that the models were not biased (*ME* null or close to 0) and had an excellent (*Rn*) to good (*FRn* and *UVA-BL*) accuracy (*RRMSE* < 20%).

The fit of the non-linear Schnute equation linking  $\zeta$  to *TrPPFD* on the calibration Dataset 1 was highly significant ( $r^2 = 0.92$ ;  $P < 0.001$ , Figure 3). Goodness-of-fit indicators showed that the model was not biased (*ME* = 0.002) and had a good accuracy (*RRMSE* = 17.9%).

The validation of the functional relationships linking *MAR* variables to *PPFD* and  $\zeta$  to *TrPPFD* on the independent Dataset 2 revealed the quality and the relevance of these relationships (Table 3; Supplementary Figure S3 & S4). Bias (*ME*) was close to 0 and very small with respect to the range of values of the four variables. The accuracy (*RRMSE*) was excellent for *Rn* and *UVA-BL*, good for *FRn*, and fair for  $\zeta$ . The linear relationships between predicted and observed data were very close to what was expected



(i.e., 1 for  $r^2$  and slopes, and 0 for y-intercepts) for  $R_n$ ,  $FR_n$  and  $UVA-BL$ , and close to what was expected for  $\zeta$  (Table 3). The lowest quality of the functional relationship for  $\zeta$  (with respect to the other variables) was related to the higher, but not biased, variability of observed  $\zeta$  around the predicted values (Figure 3), suggesting that the simplified Schnute non-linear equation with  $TrPPFD$  as predictor was appropriate, but that factor(s) other than  $TrPPFD$  contributed to  $\zeta$  variability.

### 3.2.2 Effects of positional and environmental factors on the functional relationships

#### 3.2.2.1 Linear functional relationships linking $R_n$ , $FR_n$ and $UVA-BL$ to $PPFD$

Sky condition, solar elevation and depth within the crown had a significant effect on the slope of the linear functional relationships linking  $R_n$  and  $UVA-BL$  to  $PPFD$  ( $P < 0.001$ ; Table 4). Only sky condition affected the slope of the relationship linking  $FR_n$  to  $PPFD$ , and solar elevation and depth within the crown affected the y-intercept of the relationship (Table 4).

The slope of the linear relationship linking  $MAR$  to  $PPFD$  variables was higher under clear sky than under cloudy sky for  $R_n$  and  $FR_n$ , indicating that the flux of  $R_n$  and  $FR_n$  photons increased more rapidly with  $PPFD$  under clear sky. On the contrary, the slope of the relationship was higher under cloudy sky for  $UVA-BL$ . Similarly, solar elevation and depth within the crown had opposite effects on the slope of the relationships linking  $R_n$  to  $PPFD$  and  $UVA-BL$  to  $PPFD$ . Thus, photon flux for red and blue wavelengths in the  $PPFD$  appeared to vary antagonistically. With respect to solar elevation, the slope was the highest at low solar elevation in the morning (7:00) for  $R_n$ , whereas, at that time, it was the lowest for  $UVA-BL$ . For  $R_n$  and  $UVA-BL$ , slopes were similar at high solar elevation (12:00) and at low solar elevation in the afternoon (17:00). Interestingly, the slopes were significantly different for the low solar elevation in the morning (7:00) and in the afternoon (17:00), suggesting that the slope was affected by factors other than solar elevation, differing between early morning and late afternoon. With respect to depth within the crown, the slope was higher in the middle leafy part of the crown than in the outer and inner parts for  $R_n$ , whereas the lowest slope was recorded in the middle leafy part for  $UVA-BL$ . This indicated that the flux of  $R_n$  or  $UVA-BL$  photons increased differently with  $PPFD$  in the middle leafy part of the crown, on the one hand, and in the outer and inner parts of the crown, on the other.

The y-intercept of the relationship linking  $FR_n$  to  $PPFD$  was significantly higher at high solar elevation (12:00) than at low solar elevation (7:00 and 17:00). In addition, it was significantly higher in the middle leafy part of the crown, lower in the outer part of the crown, and intermediate in the inner part of the crown (Table 4). Those results indicated that, on average, the flux of  $FR_n$  photons increased with  $PPFD$  in a similar way, regardless of solar elevation and depth within the crown (since none of these factors affected the slope), but that it was higher at high solar elevation and in the middle leafy part of the crown for any  $PPFD$ . However, the difference between significantly different y-intercepts ( $0.92$  and  $0.68 \mu\text{mol m}^{-2} \text{s}^{-1}$  for solar elevation and depth within the crown, respectively, Table 4) was very small in comparison to the range of  $FR_n$  values ( $0.10\text{--}73 \mu\text{mol m}^{-2} \text{s}^{-1}$ ).

The effects of solar elevation on the linear functional relationships linking  $MAR$  variables to  $PPFD$  changed with the sky condition (Table 5). Under clear sky, in accordance with previous results (Table 4), the slope was significantly higher for  $R_n$  and significantly lower for  $UVA-BL$  at low solar elevation in the morning (7:00) than at high solar elevation (12:00) and at low solar elevation in the afternoon (17:00). Under cloudy sky, the way solar elevation affected slopes of the relationships linking  $R_n$  and  $UVA-BL$  to  $PPFD$  was reversed from the clear sky condition: the slope was significantly lower for  $R_n$  and significantly higher for  $UVA-BL$  at low solar elevation in the morning than at high solar elevation and low solar elevation in the afternoon. The absolute difference between the highest and lowest slope values for  $R_n$  or  $UVA-BL$  was approximately two to three times greater under clear sky than under cloudy sky. This probably explained why the results of the global solar elevation analysis (Table 4) were similar to those obtained under clear sky (Table 5).

In accordance with previous results (Table 4), solar elevation had no (clear sky) or only marginal (cloudy sky) effect on the slope of the relationship linking  $FR_n$  to  $PPFD$ , but had a significant effect on its y-intercept. Under clear sky, the y-intercept was higher at 12:00 than at 7:00 and 17:00. Under cloudy sky, the y-intercept was significantly lower at 7:00 than at 12:00 and 17:00.

#### 3.2.2.2 Non-linear functional relationship linking $\zeta$ to $TrPPFD$

LRT analyses showed that sky condition, solar elevation and depth within the crown significantly affected the non-linear functional relationship linking  $\zeta$  to  $TrPPFD$  with the Schnute equation (Table 6; Figure 4).

TABLE 2 Parameter estimates (standard error, s.e.) of the linear models ( $y = a + b \text{PPFD}$ ; Equation 1) linking  $MAR$  variables ( $R_n$ ,  $FR_n$ ,  $UVA-BL$ ) to photosynthetic photon flux density ( $PPFD$ ) in mango tree crowns, and statistics of model calibration and goodness-of-fit ( $P$ ,  $r^2$ :  $P$ -value and coefficient of determination of the regression line between  $PPFD$  and  $MAR$  variables;  $ME$ , mean error;  $RRMSE$ , percentage relative root mean square error).

MAR variables	$a$		$b$		$P$	$r^2$	$ME$	$RRMSE$
	Estimate	(s.e.)	Estimate	(s.e.)				
$R_n$	0	–	0.035	(2.77e-05)	< 0.001	0.998	0.12	7.3
$FR_n$	0.73	(0.034)	0.032	(8.02e-05)	< 0.001	0.999	0	18.5
$UVA-BL$	1.84	(0.211)	0.338	(5.06e-04)	< 0.001	0.995	0	12.3

**TABLE 3** Statistics of model validation performed on the independent Dataset 2 for the linear models (Equation 1) predicting narrowband red (*Rn*), narrowband far-red (*FRn*) and UVA-blue (*UVA-BL*) as a function of the photosynthetic photon flux density (*PPFD*), and for the simplified Schnute non-linear equation (Equation 3) predicting the *Rn* : *FRn* ratio ( $\zeta$ ) as a function of the relative transmitted *PPFD* (*TrPPFD*) in mango tree crowns (*Slope*, *y-intercept*,  $r^2$ : slope, y-intercept and coefficient of determination of the regression line between predicted and observed values; 95% CI: 95% confidence interval around the slope and y-intercept estimates; *ME*, mean error; *RRMSE*, percentage relative root mean square error).

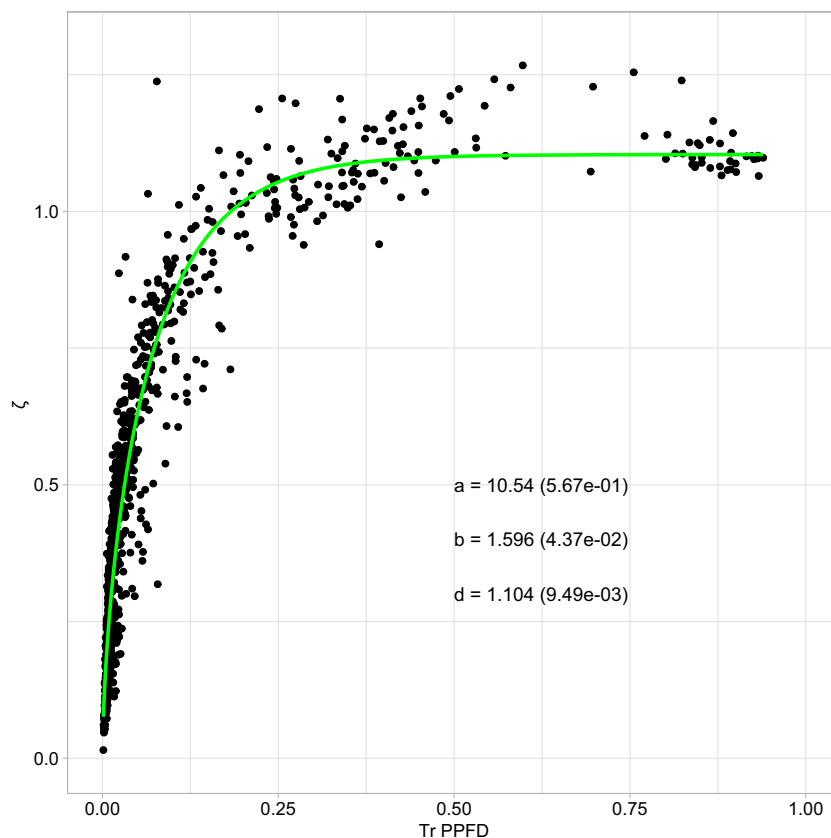
MAR variable		<i>y</i> -intercept Estimate	[95% CI]	<i>Slope</i> Estimate	[95% CI]	$r^2$	<i>ME</i>	<i>RRMSE</i>
<i>Rn</i>	(Equation 1)	0.134	[0.108; 0.160]	0.999	[0.998; 1.001]	1	0.129	4.9
<i>FRn</i>	(Equation 1)	-0.500	[-0.602; -0.397]	1.030	[1.024; 1.036]	0.99	-0.280	17.8
<i>UVA-BL</i>	(Equation 1)	0.636	[0.150; 1.121]	0.996	[0.994; 0.999]	1	0.383	8.4
$\zeta$	(Equation 3)	0.084	[0.073; 0.094]	0.888	[0.870; 0.906]	0.94	0.032	21.2

With respect to sky condition,  $\zeta$  was higher under cloudy sky than under clear sky, in particular when *TrPPFD*  $\leq$  0.30 (Figures 4A, B). The  $\zeta$  difference between the two sky conditions reached up to 0.1 for small *TrPPFD* (0.02).

The functional relationships estimated for the three solar elevations were significantly different from each other (Table 6; Figures 4C, D). In particular, the relationships at low solar elevation in the morning (7:00) and in the afternoon (17:00) were significantly different, suggesting that factor(s) other than solar elevation affected the relationship. When *TrPPFD*  $\leq$  0.30 (except for

very small *TrPPFD* values),  $\zeta$  was lower at low solar elevation in the morning (7:00) and in the afternoon (17:00) than at high solar elevation (12:00). The maximal difference was 0.08 at *TrPPFD* = 0.10, and 0.07 at *TrPPFD* = 0.14 for the 7:00 and 17:00 levels, respectively. When *TrPPFD* > 0.30,  $\zeta$  became higher at low solar elevation in the morning (7:00) than at high solar elevation (12:00), whereas at low solar elevation in the afternoon (17:00), it remained lower than at high solar elevation (12:00).

The functional relationships estimated for the three depths within the crown at the highest solar elevation (12:00) were also



**FIGURE 3** Non-linear functional relationship linking the MAR variable  $\zeta$  (*Rn* : *FRn* ratio) to the relative transmitted photosynthetic photon flux density (*TrPPFD*) in mango tree crowns. Points are observed data and the green line represents the simplified Schnute equation,  $\zeta = d \left[ \frac{1 - e^{-a \cdot TrPPFD}}{1 - e^{-a}} \right]^{1/b}$  (Equation 3), fitted to data, with the estimates ( $\pm$  standard error, s.e.) of model parameters indicated on the graph.

**TABLE 4** Slope estimates and *F*- and *P*-values of the analyses of covariance testing the effects of sky condition, solar elevation and depth within the crown on the slope of the linear functional relationships linking narrowband red (*Rn*), narrowband far-red (*FRn*), and UVA-blue (*UVA-BL*) to photosynthetic photon flux density (*PPFD*) in mango tree crowns.

Factor	Level	<i>Rn</i>	<i>FRn</i> <sup>(2)</sup>	<i>UVA BL</i>
Sky condition	Clear	0.0353 $\alpha$	0.0315 $\alpha$	0.337 $\beta$
	Cloudy	0.0333 $\beta$	0.0288 $\beta$	0.373 $\alpha$
	<i>F</i> <sub>1,860</sub>	239.1	48.0	211.4
	<i>P</i>	<b>&lt; 0.001</b>	<b>&lt; 0.001</b>	<b>&lt; 0.001</b>
Solar elevation	7:00	0.0376 $\alpha$	0.0313	0.292 $\beta$
	12:00	0.0353 $\beta$	0.0312	0.339 $\alpha$
	17:00	0.0347 $\beta$	0.0298	0.349 $\alpha$
	<i>F</i> <sub>2,858</sub>	74.8	2.5	82.3
	<i>P</i>	<b>&lt; 0.001</b>	0.087	<b>&lt; 0.001</b>
Depth within the crown <sup>(1)</sup>	Outer	0.0353 $\beta$	0.0314	0.337 $\alpha$
	Middle	0.0357 $\alpha$	0.0314	0.330 $\beta$
	Inner	0.0352 $\beta$	0.0315	0.340 $\alpha$
	<i>F</i> <sub>2,282</sub>	14.6	0.04	16.4
	<i>P</i>	<b>&lt; 0.001</b>	0.96	<b>&lt; 0.001</b>

<sup>(1)</sup>: analysis was performed with the data subset corresponding to the high solar elevation (12:00).

<sup>(2)</sup>: when the factor had no significant effect on the slope, its effect was tested on the y-intercept of the linear relationship. Solar elevation: *F*<sub>2,860</sub> = 93.3; *P* < 0.001; y-intercept = 0.442  $\beta$  (7:00), 1.36  $\alpha$  (12:00) and 0.520  $\beta$  (17:00). Depth within the crown: *F*<sub>2,284</sub> = 6.5; *P* = 0.002; y-intercept = 0.881  $\beta$  (outer), 1.56  $\alpha$  (middle) and 1.44  $\alpha\beta$  (inner).

For each variable and factor, slopes with the same Greek letter are not significantly different at *P* = 0.01 (pairwise comparisons with Bonferroni *P*-value adjustment). *P*-values in bold are significant at *P* = 0.01.

significantly different from each other (Table 6). The relationships estimated for the middle leafy part and for the inner part of the crown varied similarly with respect to the relationship estimated for the outer part of the crown and considered as a reference (Figures 4E, F). Values of  $\zeta$  for the middle and inner parts of the crown were higher than for the outer part of the crown when

*TrPPFD* < 0.20, with a maximum difference of 0.17 between the outer and inner parts, occurring at *TrPPFD* = 0.05. When *TrPPFD* > 0.20,  $\zeta$  for the middle and inner parts of the crown were lower than for the outer part of the crown. Over the whole *TrPPFD* range,  $\zeta$  differences with the outer part of the crown were higher for the inner part of the crown than for the middle leafy part of the crown.

**TABLE 5** Slope estimates and *F*- and *P*-values of the analyses of covariance testing the effect of solar elevation under clear or cloudy sky condition on the slope of the linear functional relationships linking narrowband red (*Rn*), narrowband far-red (*FRn*), and UVA-blue (*UVA-BL*) to photosynthetic photon flux density (*PPFD*) in mango tree crowns.

Sky condition	Solar elevation	<i>Rn</i>	<i>FRn</i> <sup>(1)</sup>	<i>UVA BL</i>
Clear	7:00	0.0382 $\alpha$	0.0310	0.282 $\beta$
	12:00	0.0353 $\beta$	0.0307	0.338 $\alpha$
	17:00	0.0359 $\beta$	0.0300	0.328 $\alpha$
	<i>F</i> <sub>2,426</sub>	72.0	0.66	82.3
	<i>P</i>	<b>&lt; 0.001</b>	0.515	<b>&lt; 0.001</b>
Cloudy	7:00	0.0323 $\beta$	0.0268	0.399 $\alpha$
	12:00	0.0333 $\alpha$	0.0279	0.373 $\beta$
	17:00	0.0332 $\alpha$	0.0294	0.375 $\beta$
	<i>F</i> <sub>2,426</sub>	9.3	5.3	16.1
	<i>P</i>	<b>&lt; 0.001</b>	0.010	<b>&lt; 0.001</b>

<sup>(1)</sup>: when solar elevation had no significant effect on the slope, its effect was tested on the y-intercept of the linear relationship. Solar elevation under clear sky: *F*<sub>2,428</sub> = 264.2; *P* < 0.001; y-intercept = 0.792  $\beta$  (7:00), 2.51  $\alpha$  (12:00) and 0.516  $\gamma$  (17:00). Solar elevation under cloudy sky: *F*<sub>2,428</sub> = 81.7; *P* < 0.001; y-intercept = 0.193  $\beta$  (7:00), 0.777  $\alpha$  (12:00), 0.693  $\alpha$  (17:00).

For each sky condition and variable, slopes with the same Greek letter are not significantly different at *P* = 0.01 (pairwise comparisons with Bonferroni *P*-value adjustment). *P*-values in bold are significant at *P* = 0.01.

**TABLE 6** Results of the LRT analysis testing the effects of sky condition, solar elevation and depth within the crown on the non-linear functional relationship linking  $\zeta$  (ratio between the narrowband red  $R_n$  and narrowband far-red  $FR_n$ ) to the relative transmitted photosynthetic photon flux density ( $TrPPFD$ ) with the simplified Schnute equation,  $\zeta = d \left[ \frac{1 - e^{-a TrPPFD}}{1 - e^{-a}} \right]^{1/b}$  (Equation 3), in mango tree crowns.

Factor	Model	Levels		Parameters			RSS	N	LRT	df	P
				a	b	d					
Sky condition	$\omega$			10.54	1.596	1.104	7.831	864	169.0	3	< <b>0.001</b>
	$\Omega$	Clear	$\alpha$	9.13	1.502	1.113	4.751				
		Cloudy	$\beta$	8.64	1.973	1.121	1.689				
Solar elevation	$\omega$			10.54	1.596	1.104	7.831	864	39.0	6	< <b>0.001</b>
	$\Omega$	7:00	$\alpha$	8.53	1.676	1.140	2.893				
		12:00	$\beta$	14.63	1.321	1.100	1.643				
		17:00	$\gamma$	10.78	1.712	1.062	2.949				
Depth crown <sup>(1)</sup>	$\omega$			14.63	1.32	1.10	1.643	288	76.7	6	< <b>0.001</b>
	$\Omega$	Outer	$\alpha$	13.54	1.129	1.115	0.432				
		Middle	$\beta$	16.37	1.240	1.071	0.439				
		Inner	$\gamma$	26.87	1.045	1.021	0.388				

<sup>(1)</sup>: analysis was performed with the data subset corresponding to the high solar elevation (12:00).

Values are parameter estimates and residual sums of squares (RSS) for the coincident model ( $\omega$ ) and the base models ( $\Omega$ ) estimated for each factor level and outputs of LRT tests comparing the coincident and base models for each factor (N: sample size; LRT: LRT statistics; df: number of degrees of freedom; and P: P-value). For each factor, relationships for factor levels with the same Greek letter are not significantly different at  $P = 0.01$  (pairwise comparisons with Bonferroni P-value adjustment). P-values in bold are significant at  $P = 0.01$ .

As a consequence,  $\zeta$  was higher inside than at the periphery of the mango tree crown for low  $TrPPFD$ , whereas the reverse ranking of  $\zeta$  was observed for higher  $TrPPFD$ , with higher values at the periphery than inside the crown.

The non-linear functional relationship linking  $\zeta$  to  $TrPPFD$  was affected differently by solar elevation under clear or cloudy sky (Table 7). Under clear sky, the relationships estimated for each solar elevation were significantly different from each other. Under cloudy sky, the relationships estimated at low solar elevation in the morning (7:00) and at high solar elevation (12:00) were not significantly different, and both differed from the relationship estimated at low solar elevation in the afternoon (17:00). As for the previous result obtained on the complete dataset (Table 6), the relationships estimated at low solar elevation in the morning and in the afternoon were significantly different, independently of the sky condition.

## 4 Discussion

This research investigated the variability of light quantity (PAR variables:  $PPFD$ ,  $TrPPFD$ ) and quality (MAR variables:  $UVA-BL$ ,  $R_n$ ,  $FR_n$ ,  $\zeta = R_n : FR_n$ ), and the relationships between light quality and light quantity within the crown of the evergreen mango tree. We also explored how this variability and the relationships were shaped by the depth within the mango crown and affected by two environmental factors: sky condition and solar elevation. Besides the originality of the approach, the results are useful for studying and modeling light-related processes in plants (photosynthesis, photomorphogenesis).

### 4.1 Variability of light quantity and quality within the mango tree crown

The experimental design that took account of different depths within the crown, solar elevations and sky conditions made it possible to capture almost the whole range of light conditions with  $TrPPFD$  varying from 0.001 to 0.94. This range was slightly wider than the one measured in neotropical rainforests (0.01–0.9; Lee, 1987). Similarly, the range of measured  $\zeta$  values was large (0.02–1.27) and covered almost the whole range of this variable. The lowest  $\zeta$  values in the mango tree crown were closer to the ones measured in temperate beech and spruce forests (0.02–0.05; Ammer, 2003; Hertel et al., 2011) or wheat stands (0.05; Evers et al., 2007a) than to those measured in neotropical rainforests (0.17; Lee, 1987). These results indicated that because of the evergreen status of the species, the whole range of light quantity and quality can be found permanently within mango tree crown, with negative consequences on photosynthesis, bud burst, and modifications of axis morphology in the most shaded parts of the crown.

The variability of all variables related to light quantity and quality was highly structured by the depth within the crown (Table 1), with a significant decrease of the variables with increasing depth (Figure 2). However, this decrease was not regular, with a larger decrease between the outer and the middle part of the crown than between the middle and the inner part of the crown, probably in relation to the strong interactions between light and leaves in the dense, leafy middle part of the crown. Light quality and quantity were indeed less modified from the middle to the inner part of the crown where leaves were scarce and scattered and, consequently, where interactions were limited.  $FR_n$  was on average

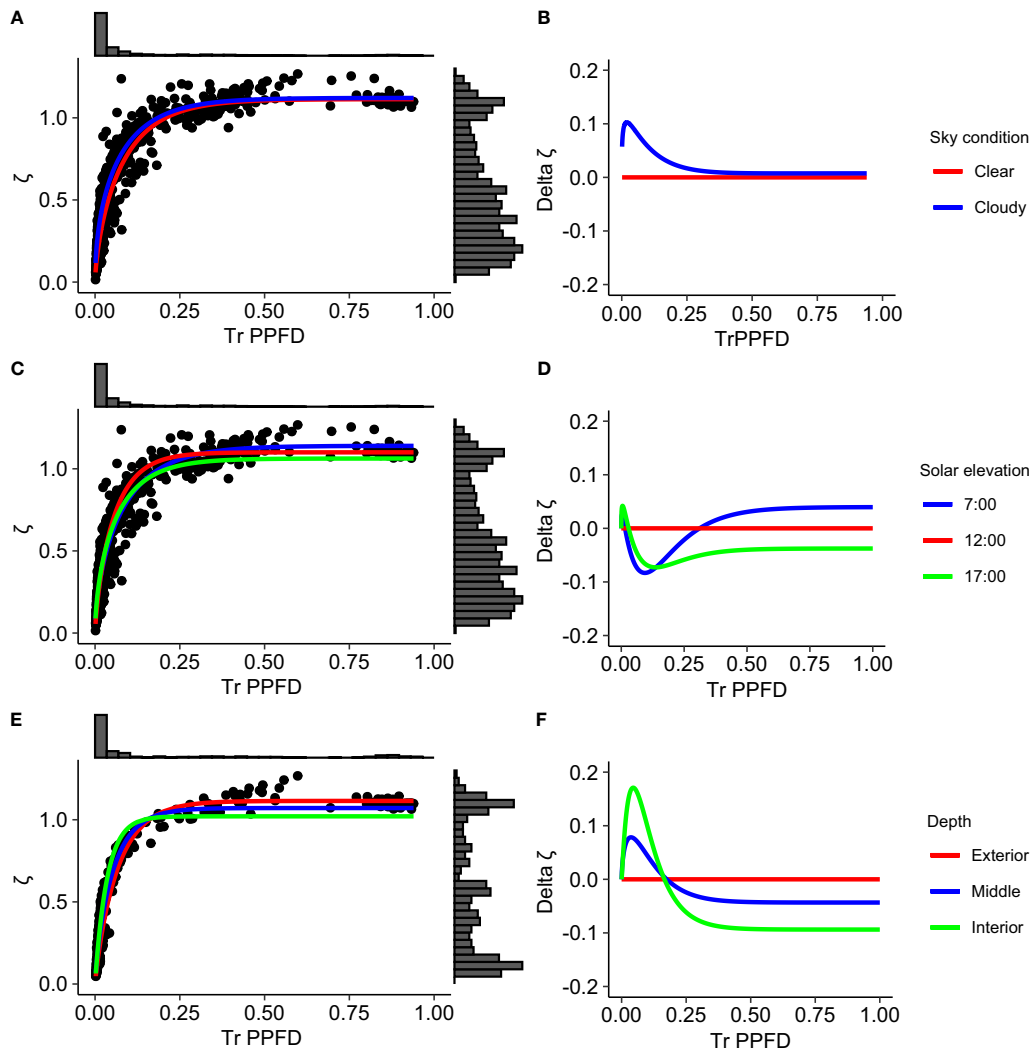


FIGURE 4

Non-linear functional relationships linking the *MAR* variable  $\zeta$  ( $Rn : FRn$  ratio) to the relative transmitted photosynthetic photon flux density ( $TrPPFD$ ) in mango tree crowns, according to sky condition (A, B), solar elevation (C, D) and depth within the crown (E, F). Graphs (A, C, E) points are observed data, curves represent the simplified Schnute equation,  $\zeta = d \left[ \frac{1 - e^{-a TrPPFD}}{1 - e^{-a}} \right]^{1/b}$  (Equation 3), fitted to data for each factor level, and histograms are distributions of  $\zeta$  and  $TrPPFD$  variables. Graphs (B, D, F) curves represent the difference (delta  $\zeta$ ) between  $\zeta$  predicted by the model for each factor level and  $\zeta$  predicted for a reference level as a function of  $TrPPFD$ . 'Clear', '12:00' and 'Outer' are the reference levels for sky condition, solar elevation and depth within the canopy factors, respectively.

less attenuated that  $Rn$  and  $UVA-BL$  between the outer and the inner parts of the crown (79% vs. 84%, respectively), in relation to the preferential absorption of  $UVA-BL$  and  $Rn$  compared to  $FRn$  by the mango leaves (Lee et al., 1987). These authors measured that average absorption of  $UVA-BL$ ,  $Rn$  and  $FRn$  by mango leaves of the cultivar Tommy Atkins was 94%, 91% and 13%, respectively.  $FRn$  is mostly reflected or transmitted, through scattering on leaf surface according to the structure and orientation of the leaf (Ustin and Jacquemoud, 2020). The differential absorption of  $Rn$  and  $FRn$  explained the decrease of  $\zeta$  within the crown. These results were similar to data recorded in temperate forests (Hertel et al., 2011) or individual peach trees (Baraldi et al., 1994) despite different depth gradients (radial/horizontal in our experiment and vertical in published studies), suggesting that the same physical processes of interactions between light and phytoelements are involved.

Light quantity and quality variables in the mango tree crown were lower under cloudy sky than under clear sky, except  $\zeta$ , which showed an opposite trend. These results are in agreement with those obtained in various vegetation types and structures (Messier et al., 1989; Pecot et al., 2005; Reinhardt et al., 2010; Hertel et al., 2011; Dengel et al., 2015). Clouds modify both the quantity and quality of incident solar light (Durand et al., 2021). They decrease the overall quantity of light and increase the diffuse fraction of light because of scattering. Solar light is enriched in blue wavelengths and depleted in red and far-red wavelengths under cloudy sky (Durand et al., 2021; Musiolková et al., 2021). Moreover, water vapor more specifically absorbs the 728 nm band corresponding to  $FRn$ , leading to higher  $\zeta$  values in incident solar light under cloudy sky (Lee and Downum, 1991). In mango tree crown, the similar variations in average  $PPFD$ ,  $Rn$ ,  $FRn$  and  $UVA-BL$  values between clear and cloudy skies (Figure 2) suggested that the decrease in



**TABLE 7** Results of the LRT analysis testing the effects of solar elevation under clear or cloudy sky condition on the non-linear functional relationship linking  $\zeta$  (ratio between the narrowband red  $R_n$  and narrowband far-red  $FR_n$ ) to the relative transmitted photosynthetic photon flux density ( $TrPPFD$ ) with the simplified Schnute equation,  $\zeta = d \left[ \frac{1 - e^{-a TrPPFD}}{1 - e^{-a}} \right]^{1/b}$  (Equation 3), in mango tree crowns.

Sky condition	Model	Solar elevation		Parameters			RSS	N	LRT	df	P
				a	b	d					
Clear	$\omega$			9.13	1.502	1.113	4.751	432	71.3	6	<b>&lt;0.001</b>
	$\Omega$	7:00	$\alpha$	6.20	1.621	1.185	1.483				
		12:00	$\beta$	11.10	1.185	1.099	0.280				
		17:00	$\gamma$	9.95	1.708	1.089	4.028				
Cloudy	$\omega$			8.64	1.973	1.121	1.689	432	51.6	6	<b>&lt;0.001</b>
	$\Omega$	7:00	$\alpha$	8.54	1.994	1.127	0.496				
		12:00	$\alpha$	8.03	2.085	1.155	0.333				
		17:00	$\beta$	10.03	1.815	1.064	0.670				

Values are parameter estimates and residual sums of squares (RSS) for the coincident model ( $\omega$ ) and the base models ( $\Omega$ ) estimated for each factor level and outputs of LRT tests comparing the coincident and base models for each factor (N: sample size; LRT: LRT statistics; df: number of degrees of freedom; and P: P-value). For each factor, relationships for factor levels with the same Greek letter are not significantly different at  $P = 0.01$  (pairwise comparisons with Bonferroni P-value adjustment). P-values in bold are significant at  $P = 0.01$ .

$R_n$ ,  $FR_n$  and  $UVA-BL$  under cloudy sky was predominantly explained by the decrease in  $PPFD$  since these variables were linearly related (see below), as shown in Musiolková et al. (2021). In contrast, the higher average  $\zeta$  values under cloudy sky, irrespective of the depth within the crown (Figure 2E), was probably related to the higher  $\zeta$  value of incident solar light and to the higher proportion of diffuse incident light that can penetrate deeper within the crown (Li and Yang, 2015). The differences were considerable since average  $\zeta$  values under cloudy sky in the middle and inner parts of the crown were about 0.5, similar to those in the outer part under clear sky in the early morning or late afternoon.

Solar elevation affects incident light quantity and quality by altering the length of solar ray paths in the atmosphere (Holmes and Smith, 1977). The results on light variability within the mango tree crown were consistent with this rule.  $PPFD$ ,  $TrPPFD$ ,  $R_n$ ,  $FR_n$  and  $UVA-BL$  were significantly affected by solar elevation, with higher average values at high solar elevation (12:00) than at low solar elevation (7:00 and 17:00). Similarly to the effect of sky condition and for the same reason, the effect of solar elevation on  $R_n$ ,  $FR_n$  and  $UVA-BL$  was probably explained by the effect of solar elevation on  $PPFD$  (Figure 2). On the contrary, the ratio  $\zeta$  was not significantly affected by solar elevation within the mango tree crown. Durand et al. (2021) reported that  $\zeta$  within canopies can be constant or vary with solar elevation according to the sky condition or to the species. For the mango tree crown, the stability of  $\zeta$  with solar elevation was related to the similar changes of  $R_n$  and  $FR_n$  with solar elevation for each sky condition (Figures 2C, D) probably in relation to their linear dependence with  $PPFD$  (Supplementary Figure S3).

### 4.2 Variability of the relationships linking light quality to light quantity within the mango tree crown

The large variability of the variables describing light quantity and light quality within a mango tree crown made it possible to

determine the functional relationships between light quality and light quantity, and, more originally, to investigate how these relationships varied within the crown and with sky condition and solar elevation.

$R_n$ ,  $FR_n$  and  $UVA-BL$  were linearly related to  $PPFD$ . The relationships were strong, unbiased, with little variability around the trend, and excellently validated with independent data. The estimated slopes were very close to those estimated for the same relationships in a sorghum stand (Escobar-Gutiérrez et al., 2009), despite considerable differences in plant material (e.g., stand architecture, leaf morphology, geometry and optical properties) and measurement protocols (e.g., sensor orientations). This suggests that the relationships for  $R_n$  and  $UVA-BL$  mostly reflected the fact that these wavelength bands are completely or almost completely included in the  $PAR$ . Variation in  $PPFD$  consequently led to proportional variation in the photon flux of these  $MAR$  bands. The strong linear relationship between  $FR_n$  and  $PPFD$  showed that there was also proportionality between  $PPFD$  and the photon flux of the  $FR_n$  wavelength band close to, but outside the  $PAR$  range. This probably explained why the goodness-of-fit of the linear relationships was slightly lower for  $FR_n$  in calibration and validation ( $RRMSE$ ; Tables 2 and 3).

Depth within the crown, sky condition and solar elevation affected the slopes of the relationships for  $R_n$  and  $UVA-BL$ , i.e., the proportion of these wavelength bands within  $PPFD$ . Only the sky condition affected the slope of the relationship for  $FR_n$ , and depth within the crown and solar elevation affected the y-intercept, i.e., generated a constant difference in  $FR_n$  flux within the  $PPFD$  range. The slopes of the relationships for  $UVA-BL$  were higher under cloudy sky than under clear sky, whereas the opposite was true for  $R_n$  and  $FR_n$ . This was probably related to the quality of incident light that is enriched in blue wavelengths and depleted in red and far-red wavelengths under cloudy sky (Navrátil et al., 2007; Durand et al., 2021; Musiolková et al., 2021). The lack of effect of the factors, solar elevation and depth within the crown, on the slope of

the relationship between  $FRn$  and  $PPFD$  suggested that the effects of these factors were related to the higher absorbance of  $Rn$  and  $UVA-BL$  than  $FRn$  by leaves (Lee et al., 1987).

In the four relationships linking  $Rn$  and  $UVA-BL$  to  $PPFD$  under two sky conditions (Table 5), slopes at high solar elevation and at low solar elevation in the afternoon were similar and differed from the slope at low solar elevation in the morning. This indicated that solar elevation was not the only factor involved in this effect since slopes of the relationships at low elevation differed between morning and afternoon. Such differences in light quality between morning and afternoon for similar solar elevation have also been recorded at two distinct geographical (Finland, India) and climatological locations (Kotilainen et al., 2020). Two other factors could be involved, but further investigations are required to decipher the effects of solar elevation on these relationships. First, air humidity is generally higher in the morning and could therefore affect the proportion of  $Rn$  and  $UVA-BL$  in incident light in comparison to the afternoon since atmospheric water vapor affects, independently of the sky condition, the relative proportion of blue and red wavelengths in incident light (Kotilainen et al., 2020). Second, photosynthesis generally occurs in the morning when the vapor pressure deficit is low and stomata are still open, and could lead to differential absorption of  $Rn$  and  $UVA-BL$  by leaves (Gitelson et al., 2021; Liu and van Iersel, 2021).

Our assumption that the effect of solar elevation on the functional relationships was more likely under the clear than cloudy sky condition was not completely supported by the results since the effect was significant under both sky conditions. However, differences in slope values for  $Rn$  and  $UVA-BL$  relationships, and in y-intercept for  $FRn$  relationships between solar elevations were larger under clear sky than under cloudy sky (Table 5), suggesting that the more isotropic directional distribution of light beams under cloudy sky mitigated the effect of solar elevation on the relationships between light quality and quantity.

The ratio  $\zeta$  was non-linearly related to  $TrPPFD$  in the mango tree crown, as revealed in various types of vegetation (Muraoka et al., 2001; Leuchner et al., 2007; Escobar-Gutiérrez et al., 2009). The relationship was strong, unbiased and satisfactorily validated with independent data. The variability around the trend was, however, larger than that observed in the previous linear relationships. The ratio  $\zeta$  remained high ( $1 <$ ) and stable for  $TrPPFD > 0.30$ , and decreased rapidly for  $TrPPFD < 0.25$ . The curvature of the relationship between  $\zeta$  and  $TrPPFD$  was higher than that observed in a sorghum stand, as supported by the different value of the shape parameter 'a' in the Schnute equation (Equation 3) used to fit the relationships in these two species, whereas the two other parameters ('b' and 'd' in Equation 3) were similar (Escobar-Gutiérrez et al., 2009). The variability in the shape of this relationship among various vegetation types (see Muraoka et al., 2001; Leuchner et al., 2007) and the variability of the data around the trend (Figure 3) suggest that the relationship between  $\zeta$  and  $TrPPFD$  is mostly shaped by physical interactions between light and phytoelements, and is therefore affected, globally and locally, by factors such as stand/canopy architecture and leaf morphology, geometry and optical properties. In particular, genotype appears as the main source of variability of leaf optical properties (Atherton

et al., 2017; Hovi et al., 2017), because of differences in leaf biochemical composition and anatomical features (Govaerts et al., 1996), notably for wavelengths above 700 nm (Hovi et al., 2017), i.e., including  $FRn$ . Consequently, the relationship modeled in this study may vary with mango cultivar, and also with the characteristics of the neighboring trees (age, dimensions, method for canopy management) or the orchard design (planting density and/or trees arrangement), which may influence the stand/canopy architecture. Measurements should be performed in various orchards to evaluate this variability.

The three factors (depth within the crown, sky condition and solar elevation) affected the non-linear relationship. The most pronounced effects of these factors were observed in low light environments, below a  $TrPPFD$  value varying between 0.20 and 0.30 depending on the factor (Figure 4). This strengthened the idea that this relationship was shaped by the interactions between light and phytoelements since these interactions led to low  $TrPPFD$ . In particular, the effect of depth within the crown was opposite of what was expected in low light environments ( $TrPPFD < 0.20$ ) since  $\zeta$  was higher in the inner part than in the outer part of the crown. The difference reached 0.17, which was considerable in relation to the low  $\zeta$  values for low  $TrPPFD$ . The expected decrease of  $\zeta$  between the outer and the inner part of the crown was observed for  $TrPPFD > 0.20$ . An increase of  $\zeta$  has been similarly observed in the lowest part of vertical profiles of light quality in forest stands (Hertel et al., 2011; Dengel et al., 2015). Hertel et al. (2011) hypothesized the role of unattenuated diffuse light penetrating from the side of the stand. The explanation probably involves diffuse and direct light as well as scattering within the crown, all of which are difficult to measure locally. A modeling approach of the interactions between light and phytoelements would be helpful to decipher the origins of this phenomenon.

The ratio  $\zeta$  was higher under cloudy sky than under clear sky in low light environments ( $TrPPFD < 0.30$ ), with a difference reaching 0.10 suggesting that incident diffuse light under cloudy sky could penetrate more evenly within the crown, leading to higher  $\zeta$  values (Leuchner et al., 2007; Hertel et al., 2011; Dengel et al., 2015).

Solar elevation affected the relationship linking  $\zeta$  to  $TrPPFD$  (Table 6), with different behaviors for the low solar elevation levels in the morning and in the afternoon with respect to the high solar elevation (Figure 4). At low solar elevation, incident light passed through crowns of neighboring trees in the orchard and the light in the shaded part of the crown was probably modified (through absorption and scattering) more strongly by crown and neighboring trees than when it arrived vertically on the crown at 12:00. Thus, for a low  $TrPPFD$ ,  $\zeta$  was lower at 7:00 and 17:00 than at 12:00. In medium and high light environments,  $\zeta$  in the morning (7:00) was higher than during the rest of the day, probably because of incident light characteristics. This increasing of  $\zeta$  could be related to the change in humidity content in the atmosphere during the day (Kotilainen et al., 2020). This seems contradictory with the lack of effect of solar elevation on  $\zeta$  in the mango tree crown (Table 1). This latter result was probably a consequence of a high proportion of data with small  $TrPPFD$  (see data distribution in Figure 4E). The study of the relationship linking  $\zeta$  to  $TrPPFD$  therefore made it possible to highlight the specific effects of solar elevation (as well as

sky condition and depth within the crown) for different light environments.

### 4.3 Applications to modeling and experiments on light-related processes

Coupling the functional relationships linking light quality to light quantity with light interception models into functional-structural plant models was previously done for a generic individual-based model of forage legumes (Louarn and Faverjon, 2018). In this model,  $\zeta$  was estimated from the computation of *TrPPFD* in order to simulate the interactions between the individuals. Such strategy of light signal simulation can allow rapid, accurate and dynamic simulation of light quality at the level of the organs within a tree crown model like V-Mango for the mango tree (Boudon et al., 2020; Vaillant et al., 2022). This can contribute to improve the performance of models for light-related processes such as photosynthesis and especially photomorphogenesis for which the simulation of *MAR* bands is necessary. A consequence of taking into account the easier simulation of light quality may be a better consideration of its effects on ecophysiological processes (e.g. stomata behavior, Buckley and Mott, 2013; Sun et al., 2014) and on developmental processes like bud burst and branching (Franklin, 2008; Leduc et al., 2014; Demotes-Mainard et al., 2016) in plant modeling. This latter point would lead to a dynamic interplay between bud burst and light, in particular in complex architectures like leafy tree crowns of evergreen species or deciduous species (during the season of vegetative growth), with expected new insights on the 3D plant structure development (Evers et al., 2007b).

Our study highlighted the effects of environmental factors, sky condition and solar elevation, and of the depth within the crown on the quantity and quality of light within a tree crown, and especially on the functional relationships between them. Light interception models take into account implicitly or explicitly the two environmental factors (they determine the incident light) to calculate light quantity at any point of a plant structure (Vos et al., 2010). Light quality can be computed from this light quantity simulation, as suggested above, with the functional relationship corresponding to the conditions (sky condition, solar elevation, depth within the crown) of the simulation in order to have a more accurate estimation, in particular for the estimation of  $\zeta$  at low *TrPPFD* (Figure 4).

From a practical point of view, the close functional relationships revealed between light quality and light quantity under natural conditions in the mango tree crown does not allow to analyze the respective effects of light quality and quantity on morphogenesis, such as bud burst. This type of investigation would require uncoupling these effects under controlled light conditions, for example with colored filters (Maaike Wubs et al., 2014). These results also provide basic knowledge for improving light environment within the mango tree crown via pruning and tree training, with expected positive impacts on processes promoting mango production, in particular branching (Capelli et al., 2016) and

photosynthesis (Urban et al., 2003). Finally, these results can be helpful in designing *in silico* experiments in order to determine where and how (duration, composition, effect) the light signal is perceived by the plant.

## 5 Conclusion

Light quantity and quality within the crown of the adult evergreen mango tree varied within a wide range similar to that of a forest, with probable consequences on light-related processes. This variability was structured by the radial gradient of depth within the crown, and by sky condition and solar elevation. These three factors also affected the close functional relationships between light quality and quantity. The functional relationships coupled with light interception models in functional-structural plant models can allow rapid and dynamic simulation of light quantity and quality at the level of the organs within a plant in order to improve the modeling of light-related ecophysiological and developmental processes.

## Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

## Author contributions

All authors designed the experiment. EC collected the data. EC, IG, and FN analyzed the data. EC wrote the first draft of the manuscript. All authors contributed to the article and approved the submitted version.

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## Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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## Supplementary material

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fevo.2023.1046332/full#supplementary-material>

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