RESEARCH PAPER

Seasonal time-course of gradients of photosynthetic capacity and mesophyll conductance to CO₂ across a beech (Fagus sylvatica L.) canopy

Pierre Montpied¹,²,* André Granier¹,² and Erwin Dreyer¹,²,*

¹ INRA, UMR 1137, Ecologie et Ecophysiologie Forestières, F-54280 Champenoux, France
² Nancy-Université, UMR 1137, Ecologie et Ecophysiologie Forestières, F-54500 Vandoeuvre les Nancy, France

Received and Revised 26 November 2008; Accepted 25 February 2009

Abstract

Leaf photosynthesis is known to acclimate to the actual irradiance received by the different layers of a canopy. This acclimation is usually described in terms of changes in leaf structure, and in photosynthetic capacity. Photosynthetic capacity is likely to be affected by mesophyll conductance to CO₂ which has seldom been assessed in tree species, and whose plasticity in response to local irradiance is still poorly known. Structural [N and chlorophyll content, leaf mass to area ratio (LMA)] and functional leaf traits [maximum carboxylation rate (Vcmax), maximum light-driven electron flux (Jmax), and mesophyll conductance (gi)] were assessed by measuring leaf response curves of net CO₂ assimilation versus intercellular CO₂ partial pressure, along a vertical profile across a beech canopy, and by fitting a version of the Farquhar model including gi. The measurements were repeated five times during a growth season to catch potential seasonal variation. Irradiance gradients resulted in large decreasing gradients of LMA, gi, Vcmax, and Jmax. Relative allocation of leaf N to the different photosynthetic processes was only slightly affected by local irradiance. Seasonal changes after leaf expansion and before induction of leaf senescence were only minor. Structural equation modelling confirmed that LMA was the main driving force for changes in photosynthetic traits, with only a minor contribution of leaf Nitrogen content. In conclusion, mesophyll conductance to CO₂ displays a large plasticity that scales with photosynthetic capacity across a tree canopy, and that it is only moderately (if at all) affected by seasonal changes in the absence of significant soil water depletion.

Key words: Canopy, European beech, irradiance, leaf structure, LMA, mesophyll conductance, photosynthetic capacity, plasticity.

Introduction

During the last decades, the increase of atmospheric CO₂ concentration and its consequences on climate have promoted a growing interest in the function of forest ecosystems as potential carbon sinks. This has led to the development of a number of functional models aiming at predicting the future role and importance of forests in carbon assimilation and storage. One of the key steps in these models is the estimation of carbon assimilation by leaves and its integration to canopy scale by considering the vertical gradients of environmental variables and of leaf photosynthetic capacities within canopies (Baldocchi and Harley, 1995; Chen et al., 1999). Most of them use the model of Farquhar et al. (1980) for estimating leaf photosynthesis, as this approach has the advantage of including leaf photosynthetic capacity as described by the maximal carboxylation rate (Vcmax) and maximal light-driven electron flux (Jmax). Photosynthetic capacity is independent of instantaneous climate fluctuations, expresses potential rather than actual photosynthesis, and is plastic, i.e. may change with micro-climate over days or weeks. Recently, several studies have pointed out the need to include seasonal variations of leaf photosynthetic capacity
in such models (Wilson et al., 2000, 2001; Wang et al., 2003; Muraoka and Koizumi, 2005; Kosugi and Matsuo, 2006). These studies, however, focused on sun leaves or simply distinguished between shade and sun leaves but did not take into account the vertical irradiance gradient as a continuum under which plasticity of leaf structure and photosynthetic capacity is expressed, nor did they consider potential seasonal changes of leaf traits along the gradient.

They also investigated only apparent values of the parameters ($J_{\text{max}}$ and, if not only, $V_{\text{max}}$) of the model of Farquhar et al. (1980). Indeed, these parameters are usually derived from curves of net CO$_2$ assimilation rate ($A$) versus intercellular ($C_i$) and not chloroplastic CO$_2$ partial pressure ($C_c$). In other words, they assume that mesophyll conductance ($g_i$) to CO$_2$ is infinite. The use of apparent values of $V_{\text{max}}$ and $J_{\text{max}}$ is not problematic when the aim is to predict net CO$_2$ assimilation rates. It might be a serious drawback when analysing photosynthetic capacity in relation to leaf nitrogen (Ethier and Livingston, 2004) as apparent photosynthetic capacity and especially apparent $V_{\text{max}}$ are largely underestimated compared with their actual values. Recent investigations revealed that $g_i$ was smaller in shade than in sun leaves (Piel et al., 2002; Flexas et al., 2008). Further, low $g_i$ might severely limit the rate of actual net CO$_2$ assimilation rate ($A$). The gradient of $g_i$, if any, and its seasonal variations therefore need to be documented.

There are basically three techniques to estimate $g_i$: online $^{13}$CO$_2$ discrimination (Evans et al., 1986), coupled measurements of chlorophyll fluorescence and gas exchange (Di Marco et al., 1990; Loreto et al., 1992), and derivation from $A$–$C_i$ curves (Ethier and Livingston, 2004). A detailed account of these techniques can be found in Warren (2006). The first two techniques require complex measurements that might be difficult to deploy under field conditions, while the third one only requires accurate measurements of $A$–$C_i$ relationships from which actual $V_{\text{max}}$, $J_{\text{max}}$ and $g_i$ can be derived.

Another important issue for our understanding of the extent and seasonal variation of the photosynthesis gradient in forest canopies is to identify the primary driver for the plastic response of photosynthetic capacity to local irradiance. Irradiance-elicted plasticity of leaf traits has been the subject of a large number of investigations (Ellsworth and Reich, 1993; Niinemets and Tenhunen, 1997; Davi et al., 2008), and a general result was that decreased irradiance resulted in decreased leaf mass to area ratio (LMA), decreased leaf thickness (fewer layers of palisade parenchyma cells), almost constant nitrogen content, and increased chlorophyll content on a mass basis. The main question therefore is to check whether the observed plasticity in photosynthesis (as described by $V_{\text{max}}$, $J_{\text{max}}$, $g_i$, and chlorophyll content) is directly driven by changes in LMA, by changes in $N_{\text{mass}}$ (nitrogen content on a mass basis), or by changes in the relative allocation of nitrogen to photosynthesis. This question may be addressed using path analysis (Shipley, 2000).

The objectives of this study were to answer the following questions. (i) Is the irradiance-induced gradient of photosynthetic capacity paralleled by a gradient in mesophyll conductance to CO$_2$ ($g_i$) across a beech canopy? (ii) Is there a seasonal variation of photosynthetic capacity and of the irradiance-induced gradient across the canopy? (iii) Is the LMA the primary driving force for the plastic responses of photosynthesis across the canopy?

To answer these questions, $A$–$C_i$ responses curves were recorded in situ, on leaves across a beech canopy at Hesse, Eastern France. A detailed version (Ethier and Livingston, 2004) of the Farquhar et al., (1980) model was fitted to these curves to derive actual values of $V_{\text{max}}$, $J_{\text{max}}$ and $g_i$. Measurements were repeated five times during the growth season (from June to October). All data were adjusted to a common temperature of 25 °C from earlier estimates of temperature responses of the three parameters recorded on beech seedlings in a climate chamber (Dreyer et al., 2001).

### Materials and methods

#### Site

Measurements were made in a beech stand at Hesse, Northern Eastern France (48°40′N, 7°04′E, 305 m a.s.l.). The site is located amidst a 65 ha homogenous stand of 35-year-old beech trees. Tree density was ~3000 trees ha$^{-1}$ and dominant tree height ~17 m. Mean annual (1974–2003) precipitation is 860 mm and mean annual temperature is 9.3 °C (Longdoz et al., 2008). During 2001, annual precipitation was 1145 mm and rain events were evenly distributed all year round including summer (Fig. 1B); temperatures >30 °C occurred during only 4 d in August with an absolute maximum of 31.6 °C (Fig. 1C). Owing to these mild conditions, only a short and slight drought episode, i.e. with relative extractable soil water content (REW) <0.4, occurred between 12 August and 5 September (DOY 224–248; Fig. 1D). According to the course of relative global irradiance on the ground (Fig. 1A), the leaf area index (LAI) reached its maximum on 18 May and remained stable until the beginning of leaf shedding on 20 October.

#### Sampling

A scaffolding tower with floors every 2 m provided access to the canopy of ~10 trees. Five measurement campaigns were run during summer 2001 on the following dates: 19–22 June, 17–20 July, 7–10 August, 28–30 August, 4–5 October, referred to in the following as Jun, Jul, Aug1, Aug2, and Oct, respectively. Each campaign consisted of establishing a full net CO$_2$ assimilation rate—intercellular CO$_2$ partial pressure ($A$–$C_i$) response curve on about 3–4 leaves per floor in the tower. As a result each campaign provided 21 curves (18 during October).

#### Gas exchange measurements

Leaf gas exchange was measured with two inter-calibrated portable gas exchange systems Li6400 (LiCor, Lincoln, NE, USA) with a 2×3 cm leaf chamber and a red–blue diode
illuminator. Conditions in the chamber were set at 25 °C (except when ambient temperature was either too high or too low; see below for standardization procedures), photosynthetic photon flux density (PPFD) at 1000 μmol m⁻² s⁻¹ (for shade leaves) or 1500 μmol m⁻² s⁻¹ (sun leaves). Each response curve was preceded by at least 30 min activation, CO₂ partial pressure in the chamber (Cₐ) was adjusted stepwise at 180, 150, 125, 100, 75, 50, 40, 30, 20, 15, 10, and 5 Pa (1 Pa≈10 vpm). Each step comprised 3 min for adjustment and stabilization followed by three data logs at 30 s intervals.

Leaves were then severed and chlorophyll content was recorded with a Minolta SPAD 502 chlorophyll-meter (Minolta Cameras Ltd, Japan). Leaves were then brought to the laboratory where leaf area was recorded with a DeltaT area meter (DeltaT Devices, Burwell, UK), and then they were dried during 48 h at 65 °C, weighed, and ground for N analysis with a CNS elemental analyser NCS2500 (CE Instruments, Thermo Quest, Italy). The following variables were computed for each leaf: LMA (g m⁻²), Nmax (g kg⁻¹), and chlorophyll content (SPAD units). LMA was used as a surrogate for the irradiance received by leaves during their growth (Niinemets and Tenhunen, 1997).

Parameter estimation

A–Cᵢ curves were fitted according to the widely used model of Farquhar and von Caemmerer (Farquhar et al., 1980) modified according to von Caemmerer (2000) and Ethier and Livingston (2004) to compute maximal carboxylation rate (Vcmax), maximal light-driven electron flux (Jmax), and dark respiration (Rd), and to take into account the mesophyll conductance to CO₂ (gᵢ). The Vcmax and Jmax estimates provided here are therefore based on chloroplastic CO₂ partial pressure (Cᵢ) and not on Cᵢ, in contrast to many published data. The general model has been described in numerous papers and the version used here can be found in Ethier et al., (2006). The parameter set provided by Ethier et al. (2006) was used and the model was fitted jointly to the Rubisco-limited and the RuBP-limited segments of the A–Cᵢ curves as described in Dreyer et al. (2001) and in Dubois et al. (2007). The estimated parameters Vcmax, Jmax, gᵢ, and Rd were obtained from non-linear least-squares regression by means of the nls function from the R 2.5.1 software. (R-Development-Core-Team, 2007). As outlined by Warren and Adams (2006), estimating gᵢ from A–Cᵢ curves might be delicate, and high quality data are required (Pons et al., 2009). If these conditions are fulfilled, such estimates of gᵢ are in agreement with independent measurements such as combined gas exchange-chlorophyll a fluorescence measurements (Niinemets et al., 2006; Warren and Dreyer, 2006). A–Cᵢ curves that were noisy (for most of them, convergence problems occurred during the fit) were therefore discarded, and the remainder were kept when the standard error of the gᵢ estimate was <50% of the estimated value. A total of 73 A–Cᵢ curves were finally retained among 97. Rd estimates based on A–Cᵢ curves were not considered reliable because of high standard errors and were not analysed further.

Temperature corrections

Most A–Cᵢ curves were measured at 25 °C leaf temperature, but in some cases external temperatures were too low to allow measurements at 25 °C. In such cases, the parameters of interest (gᵢ, Vcmax, and Jmax) were standardized later on to their value at 25 °C. To establish the parameters for the temperature response of the three variables, the data set published by Dreyer et al. (2001) for beech seedlings under a range of leaf temperatures was used. The A–Cᵢ response curves (five different leaves under the different levels of temperature) were fitted with the above-described procedure to obtain estimates of gᵢ, Vcmax, and Jmax of the same leaves at 10, 18, 25, 32, 36, and 40 °C, with five different leaves. These data were then fitted to two temperature response models.

(i) The Arrhenius function when the parameter increased monotonously with temperature (Leuning, 1997):

\[ P(T) = P(T_{ref}) \times e^{\frac{V_{ref}}{RT_{ref}}(1 - \frac{T_{ref}}{T})} \]  

where \( P(T) \) is the value of the parameter at leaf temperature \( T \) (K), \( P(T_{ref}) \) is the value of the parameter at reference
temperature $T_{\text{ref}}$ (298.15 K), $\Delta H_a$ (kJ mol$^{-1}$) is the activation energy of the given parameter, and $R$ is the gas constant (8.314 J K$^{-1}$ mol$^{-1}$);

(ii) The simplified model of Janke et al. (2004) when there was evidence of a maximum, i.e. an optimum, in temperature response:

$$P(t) = P(t_{\text{opt}}) \times e^{-\frac{(t-t_{\text{opt}})^2}{\sigma}}$$

where $P(t)$ is the value of the parameter at leaf temperature $t$ (°C), $P(t_{\text{opt}})$ is the value at the optimal temperature $t_{\text{opt}}$ (°C), and $\Omega$ (°C) is an adjustment parameter.

The fitting procedure was similar to the one used by Dreyer et al. (2001), i.e. the temperature response parameters were considered to be common to all leaves while the values of $g_i$, $V_{c_{\text{max}}}$, and $J_{\text{max}}$ at 25 °C were assumed to be variable. The overall quality of those $A$–$C_i$ curves was satisfactory and all were retained. Estimated values were weighted by the reciprocal of their squared standard error (Murtough, 2007) to minimize the influence of poorly estimated parameters.

**Statistics**

The statistical analyses were performed as covariance analyses of the effect of the measurement campaigns (date effect) with LMA (surrogate of the irradiance received by the leaves) as covariate, using the R software (R-Development-Core-Team, 2007).

The general approach was first to fit a main model of the form:

$$R_i = a_i + b_i \times \text{LMA}_{ir} + e_{ir}$$

or:

$$R_i = a_i \times \text{LMA}_{ir}^{b_{ir}} + e_{ir}$$

with indices $i$ referring to date number, $r$ to replicates; parameters $a_i$, intercept (model 3) or scaling parameter (model 4) at date $i$; $b_i$, slope (model 3) or exponent (model 4) at date $i$; $R_{ir}$, response; and $e_{ir}$, error term. The choice between the linear model (3) and the power law model (4) was based on an analysis of the residuals against predicted plots that should show a horizontal tendency. When the response variate was an estimate from a fitting procedure, such as $V_{c_{\text{max}}}$, $J_{\text{max}}$, and $g_i$, from $A$–$C_i$ curves, it was weighed by $1/SE^2$, with $SE$ the standard error of the estimate. Then a cascade of nested models were fitted by setting to zero or equalizing the parameters that were not significantly different ($P < 0.05$) from zero or not significantly different among each other. Slope or exponent parameters were checked first. At the end of the procedure, the most parsimonious (i.e. with the lowest number of parameters) model that was not significantly different ($F$-test between main and submodel) from the main model was retained. At this stage, and if required, multiple comparison tests using the multivariate $t$ distributions of the parameters were performed on parameters of interest with the multcomp R-package (Hothorn et al., 2007). Only the results of this final model will be detailed in the text. Normality and homoscedasticity were checked graphically by plotting residuals against predicted and by normal quantile to quantile plots. When heteroscedasticity was eventually evidenced, the residual variance was modelled as a power function of predicted.

**Results**

**Temperature responses of $g_i$, $V_{c_{\text{max}}}$ and $J_{\text{max}}$ of beech saplings (Fig. 2)**

The temperature response of $V_{c_{\text{max}}}$ was monotonously increasing without inflexion within the experimental temperature range (Fig. 2A) and was therefore fitted with the simple Arrhenius model (Equation 1, Table 1). Temperature responses of $J_{\text{max}}$ and $g_i$ presented a maximum (Fig. 1B, C) estimated as 39.9 ± 2.2 °C (mean ± SE) and 26.9 ± 0.6 °C, respectively, from the June model (Equation 2, Table 1). The three adjusted models were thereafter used to standardize all data at 25 °C.

**Responses of LMA to height and date (Fig. 3)**

The range of LMA values across the canopy was very large, from ~25 at the bottom to almost 120 g m$^{-2}$ at the top of the canopy. The two-way analysis of variance (ANOVA) of LMA against height and date, and their interaction was compared with its nested one-way ANOVA against height only. LMA was log-transformed to homogenize variances. This comparison showed a lack of effect of date on LMA ($n$=73, df=22 and 45, $F$=1.138, $P$=0.347). LMA was under tight control of height ($n$=73, df=5 and 67, $F$=336, $P$<0.0001), i.e. of the irradiance gradient in the canopy. The seasonal stability of LMA, its tight correlation with height in the canopy, and previous studies on the same site showing a tight linear relationship ($R^2$ >0.95) between LMA and relative PPFD (not shown) justified the use of LMA as a surrogate of the irradiance load received by the leaves (Niinemets and Tenhunen, 1997).

**Nitrogen content in leaves (N$_{\text{mass}}$) (Fig. 4)**

N$_{\text{mass}}$ was very close to 20 g kg$^{-1}$ in all leaves and canopy positions. The linear relationship (Equation 3) between LMA and N$_{\text{mass}}$ at the different dates is shown in Fig. 4. The most parsimonious model ($n$=73, df=67, $R^2$=0.382) that did not differ significantly from the main model (3) ($F$=1.18, df=4 and 63, $P$=0.330) showed that there was a small but significant negative slope of the relationship, common to all dates, and that the intercept ranking of the dates was: Aug$^1$>Jul$^2$>Aug$^2$>Oct$^3$>Jun$^4$ [levels sharing the same letter are not significantly different ($P$ >0.05)]. N$_{\text{mass}}$ increased slightly in all leaves independently from LMA (i.e. from the canopy layer) from June to July then stabilized during summer before declining again during senescence in October (inset, Fig. 4).
Estimates of chlorophyll content per leaf area (SPAD units) (Fig. 5)

Chlorophyll content per leaf area as detected by SPAD units was linked to LMA by a power function (Equation 4). Residual variance increased with predicted and was therefore modelled as a power function of predicted. The most parsimonious model \( (n=69, \text{df}=63, R^2=0.647) \) that did not differ significantly from the main model \( (F=0.921, \text{df}=4 \text{ and } 59, P=0.458) \) revealed that there was a slight but significant and positive exponent between LMA and SPAD. The exponent was stable with dates, and the scaling parameter ranked as follows: Jul\( ^a\text{>Aug1}^{ab}\text{>Oct}^{b}\text{>Aug2}\text{>Jun}^{b} \) [levels sharing same letter are not significantly different \( P \geq 0.05 \)]. In other words, there was a seasonal dynamics in SPAD units with a small increase during early summer and a decrease during late summer (inset, Fig. 5). The changes remained moderate.
Maximal carboxylation rate ($V_{cmax}$) (Fig. 6A)

$V_{cmax}$ displayed a large range of values from 10 µmol m$^{-2}$ s$^{-1}$ to >100 µmol m$^{-2}$ s$^{-1}$. The most parsimonious power law model (Equation 4, $n=73$, $df=67$, $R^2=0.827$, weighted $R^2=0.934$) that did not differ from the main model ($F=1.61$, $df=4$ and 63, $P=0.182$) revealed a tight relationship and positive gradient between $V_{cmax}$ and LMA. This relationship was not significantly different between the three mid-summer dates July, early August, and late August. The gradient was significantly lower during the June campaign whereas it was steeper in October. $V_{cmax}$ values at similar LMA were also lower in October compared with mid-summer.

Maximal light-driven electron flux ($J_{max}$) (Fig. 6B)

$J_{max}$ displayed a large variability parallel to that of $V_{cmax}$ and ranging from 10 µmol m$^{-2}$ s$^{-1}$ to almost 200 µmol m$^{-2}$ s$^{-1}$ and was tightly related to LMA. The most parsimonious power law model (Equation 4, $n=73$, $df=67$, $R^2=0.921$, weighted $R^2=0.941$) that did not differ significantly from the main model ($F=1.79$, $df=2$ and 63, $P=0.143$) led to the following conclusions: the exponent, i.e. the gradient in the canopy, was always significantly different from 0 and positive, but was lowest in June and October, largest in late August, and intermediate in July and early August. In October, $g_i$ was lower than on the other dates in the upper canopy but similar in the lower layers.

Relative nitrogen allocation to $V_{cmax}$ and $J_{max}$ (Fig. 7)

Relative allocation of leaf nitrogen to carboxylation ($P_c$) and to light-driven electron flow ($P_b$) was calculated according to Niinemets and Tenhunen (1997).

$P_c$ was linearly related to log(LMA); and a linear model (Equation 3) was fitted to log(LMA). Examination of residuals revealed that variance was increasing with
predicted; therefore, variance was modelled as a power function of predicted. The most parsimonious model ($n=73$, df=68, $R^2=0.576$) that was not significantly different from the main model ($F=1.78$, df=5 and 63, $P=0.131$) revealed (Fig. 7A) that the gradient of $P_c$ against LMA was positive during early summer (June and July), but was reversed afterwards. The gradient remained stable from early August to October, but a downward shift of $P_c$ occurred between late August and October.

$P_b$ was linearly related to LMA, and the most parsimonious model (Equation 3, $n=73$, df = 69, $R^2=0.983$) that was not significantly different from the main model ($F=1.77$, df = 6 and 63, $P=0.120$) revealed (Fig. 7B) a constant and positive gradient of $P_b$ against LMA. Nevertheless, the intercept differed significantly and was ranked as follows: Aug$^a$<Aug$^b$<Jun$^a$=Jul$^b$=Oct$^a$ [levels sharing the same letter are not significantly different ($P \geq 0.05$)].

### Nitrogen allocation to chlorophylls (Fig. 8)

Nitrogen allocation to chlorophylls was analysed based on the ratio of SPAD units to nitrogen per unit area ($N_{area}$) as a surrogate. The relationship between this ratio and LMA obeyed a power function and was fitted to Equation 4, and residual variance modelled as a power function of predicted. The most parsimonious model ($n=69$, df=63, $R^2=0.973$) that was not significantly different from the main model ($F=1.631$, df=4 and 59, $P=0.178$) revealed an exponent common to all dates, but a significantly different scaling parameter between dates, that were ranked as follows (inset, Fig. 8): Aug$^b$<Aug$^a$<Jul$^b$<Jun$^a$=Oct$^b$ [levels sharing the same letter are not significantly different ($P \geq 0.05$)]. However, absolute differences between these parameters remained very small.

### Discussion

#### Quality and reliability of photosynthetic parameter estimation

Jointly estimating the four parameters $V_{cmax}$, $J_{max}$, $g_i$, and $R_d$ from $A-C_i$ curves is technically difficult and may lead to erroneous estimates or large standard errors of the estimation (Dubois et al., 2007) unless data quality is carefully checked. For this reason, some $A-C_i$ curves that displayed a large data scatter were discarded and all further analyses were performed by weighting the estimated parameters by the reciprocal of their quality measured by the variance of their estimation. This ensured that poorly estimated parameters did not influence the analyses too severely.

Ethier and Livingston (2004) showed that assuming an infinite mesophyll conductance leads to a large underestimation of $V_{cmax}$ and to a lesser extent of $J_{max}$. In the present study, $g_i$ was assumed to remain constant during the establishment of each $A-C_i$ curve. It was recently evidenced that $g_i$ might change rapidly under a changing microenvironment and particularly with varying $C_i$ (Flexas et al., 2007, 2008) but this point is still controversial (Tazoe et al., 2009). No model of $g_i$ response to $C_i$ has yet been proposed and it is therefore impossible to integrate short-term changes of $g_i$ into a procedure to estimate parameters of Farquhar’s model. The compromise used in the present study produced estimates of photosynthetic capacity parameters closer to the ‘true’ values than to the ‘apparent’ values as derived from a fit to the original version of the model. $g_i$ estimates produced here may therefore be interpreted as a ‘mean’ mesophyll conductance over the range of CO$_2$ partial pressure values used during the establishment of the curves.

#### Temperature responses

The estimated thermal optimum of $g_i$ was $26.9\pm0.6$ °C and agrees with earlier findings of Warren and Dreyer (2006) and Yamori et al. (2006). The peak observed in the temperature response of apparent $V_{cmax}$, when estimated assuming infinite $g_i$ (Dreyer et al., 2001), disappeared when $V_{cmax}$ was estimated on a $C_i$ basis like in the observations by Warren and Dreyer (2006). $J_{max}$, however, still peaked close to 40 °C, as its estimate is only moderately influenced by $g_i$. The use of this information helped scale all results to a common temperature of 25 °C, which is a prerequisite for the analysis of plastic responses to irradiance and to seasonal changes of photosynthetic properties.

#### Gradients in the canopy

Leaf traits related to photosynthesis are well known to display large gradients across canopies, due mainly to the intracanopy gradient of local irradiance. LMA is known to respond severely to relative irradiance, either linearly (Niinemets, 1995; Niinemets and Tenhunen, 1997) or logarithmically (Fio et al., 2005; Utsugi et al., 2005). In the latter cases, the discrepancy from the linear response was mainly induced by the lowest levels of relative irradiance. The observed relationship between LMA and height was stable in time, as found by Wilson et al. (2000) and Utsugi et al. (2005), at least in the time interval between full leaf expansion and the start of senescence. Moreover, a linear relationship was found between LMA and relative irradiance during earlier measurements in the beech canopy at
Hesse. These observations confirmed that the use of LMA as a surrogate for relative irradiance was a valid simplifying choice (Ninemets, 1995; Ninemets and Tenhunen, 1997).

N_{mass} displayed a very small gradient across the canopy (<10% between extremes): this gradient might be due to a dilution effect by non-structural carbohydrates (NSCs) in leaves (Ninemets, 1995), to thicker cell walls that display low N content (Roderick et al., 1999; Shipley et al., 2006), or to a greater relative biomass of veins and petioles in sun leaves. These findings are consistent with the literature that reports a lack of gradient (Ellsworth and Reich, 1993) or a weakly negative one (Ninemets, 1995; Iio et al., 2005) for broad-leaved canopies. As a result, N_{area} paralleled tightly the gradient of LMA across the canopy.

SPAD readings are closely correlated to chlorophyll content per leaf area (Richardson et al., 2002; van den Berg and Perkins, 2004). However, the relationship between chlorophyll meter readings and chlorophyll content might change with environmental conditions such as irradiance or leaf water status (Martínez and Guiamet, 2004). In the present study, the relationship between LMA and the ratio of SPAD readings to N_{area} was almost perfectly stable throughout the growing season. On the other hand, leaves did not experience any real drought stress during the 2001 summer. One can therefore be rather confident that SPAD units can be safely used as a surrogate of chlorophyll content throughout the growing season.

The observed gradient of chlorophyll content, i.e. larger chlorophyll content on an area basis in sun leaves compared with shade leaves, agrees with the observations made by Demarez et al. (1999) in a comparable beech canopy but contradicts the findings of Lichtenthaler et al. (2007) and Iio et al. (2005). However, the latter studies were performed with only the dominant trees and excluded suppressed trees. The present study included leaves from the latter that display very small values of LMA and SPAD and therefore expanded the range of values available across the whole canopy.

Photosynthetic capacity as described by the parameters $V_{c,max}$, $J_{max}$, and $g$, displayed a steep vertical gradient across the canopy. This is now a well-established feature for apparent $V_{c,max}$ and $J_{max}$ (Ellsworth and Reich, 1993; Ninemets and Tenhunen, 1997; Meir et al., 2002). Nevertheless, vertical profiles of $g$ were to date never characterized whereas $g_i$ is known to be usually smaller in shade than in sun leaves (Hanba et al., 2002; Piel et al., 2002; Flexas et al., 2008). This confirms the occurrence of a tight coordination between the three parameters describing leaf photosynthesis, $V_{c,max}$, $J_{max}$, and $g$, that display similar and parallel variations across the canopy, i.e. similar irradiance-elicited plasticity.

If stomatal density were constant in relation to local irradiance, the path length of passive diffusion of CO₂ from substomatal cavities to the carboxylation sites would be expected to be shorter in shade leaves that have fewer palisade parenchyma layers (Eschrich et al., 1989) and therefore increased $g$. However, Aranda et al. (2003) have shown that stomatal density decreases with decreasing irradiance in F. sylvatica leaves which might counteract the effect of the increased number of palisade parenchyma layers on diffusion path length. However, structural leaf traits correlated with LMA such as mesophyll thickness or mesophyll cell surface area have also been shown to influence $g$, positively (Terashima et al., 2001). The vertical gradient of $g_i$ might, therefore, be attributed, at least partly, to a structural plasticity of leaf anatomy as expressed by LMA. Nevertheless, it is known (Piel et al., 2002) that the diffusion component is small, at least in broadleaved deciduous species (<20%) and that the liquid phase component plays a major role. A clear identification of the correlation between liquid phase conductance and cell structure and membrane constituents is still under discussion (Uehlein et al., 2008).

The vertical gradient of relative allocation of nitrogen to carboxylation ($P_c$) or to bioenergetics ($P_b$) was weak (30–50% between extreme values) with respect to the gradients displayed by $V_{c,max}$ and $J_{max}$ (~5- to 6-fold), i.e. the changes in photosynthetic capacity across the canopy were not driven by changes in N_{mass} (see above) but mainly by leaf structure (LMA to a large extent) and to a lesser extent by the relative allocation of N to photosynthesis ($P_c$ and $P_b$). The small gradient observed for $P_c$ was negative during early summer and positive later on, while $P_b$ displayed a small and positive gradient during the whole growth season. In contrast, the gradient of nitrogen allocation to light capture, as indicated by the ratio of SPAD readings to N_{area}, was steep, ~4-fold, and negative (from low to high irradiance). The structural adjustments through LMA in response to irradiance described by Terashima et al. (2001) as a direct consequence of structural and physiological constraints on CO₂ diffusion inside the leaves are compensated through changes in nitrogen allocation to light capture, so as to maintain a large leaf absorbance even in the lowest leaves with low LMA. However, a slight gradient of chlorophyll content per unit leaf area was still observed, suggesting that nitrogen shortage occurred in the lowest leaves present on suppressed trees.

**Functional links between variables**

The bivariate approach used for analysing the responses of the different variables to the irradiance gradient cannot differentiate between their direct and indirect relationships. A multivariate approach using path analysis with the d-sep method of Shipley (2000) was therefore used to disentangle the complex relationships between variables measured across the beech canopy. Analysis was performed on the covariance matrix of the log-transformed variables centred around their mean by date to take into account the structure of the data set (Shipley, 2000). Degrees of freedom were adjusted to total observation number (69) minus number of groups (five dates) to take into account centring of variables around date means. Four observations with missing data were excluded. Multinormality of the data set was checked with Mardia’s test (Shipley, 2000). The approach of centring variables by their mean per measurement campaign implicitly assumes that seasonal variations were small, as discussed below, and had only little influence.
on the detected functional relationships. This was confirmed by the fact that at least two path models were not rejected. The above discussion already suggested that LMA was the key variable controlling the overall irradiance-elicited plasticity in photosynthetic traits and that leaf nitrogen content (Nmass) only secondarily affected this plastic response. In a first step, these two variables were identified as the roots of the path models with, however, a functional link from LMA to Nmass accounting for the dilution effect of Nmass by LMA discussed above. In a second step, the other variables, all related to leaf photosynthetic capacity, were modelled as being controlled by LMA and Nmass. At this stage, several hypothetical paths could be postulated. A first model (Fig. 9, model 1) assumed that they were all controlled independently by LMA and Nmass. This model was clearly rejected (Fisher C test: \( \chi^2 = 59.2, \text{ df} = 12, P < 0.0001 \)), which indicates that some coordination occurs between these variables independently from the gradient of local irradiance. It has been outlined above that the chlorophyll content gradient is steeper when expressed on a leaf mass basis than on a leaf area basis and that it is the reverse for photosynthetic capacity (\( V_{\text{cmax}}, J_{\text{max}}, \) and \( g_i \)), suggesting that chlorophyll content behaves, with respect to the driving variables Nmass and LMA, independently of the other variables.

Therefore, several path models that assumed independence of chlorophyll content (SPAD) from photosynthetic capacity were tested. The current knowledge about the relationships between the latter variables does not allow any particular pathway relating them to be postulated, and therefore most of the models encompassing biologically sound relationships were fitted. Among these, only two (Fig. 9, models 2 and 3) were not rejected.

In model 2, LMA controls Nmass, and chlorophylls (SPAD) are controlled independently from the other photosynthesis parameters by LMA and Nmass. In addition, there is a coordination between \( V_{\text{cmax}}, J_{\text{max}}, \) and \( g_i \) in the leaves depending on nitrogen allocation: \( V_{\text{cmax}} \) is controlled by both LMA and Nmass, then \( V_{\text{cmax}} \) controls \( g_i \) and \( J_{\text{max}} \), and \( J_{\text{max}} \) also controls \( g_i \) (if this last condition is not respected, then the model is rejected). LMA also controls \( J_{\text{max}} \) independently of the previous controls (i.e. nitrogen allocation to \( J_{\text{max}} \) relative to \( V_{\text{cmax}} \) is modified by irradiance). This model was not rejected (Fisher’s C test: \( \chi^2 = 15.6, \text{ df} = 12, P = 0.210 \)).

Model 3 is similar to model 2 but \( g_i \) acts as a control over \( V_{\text{cmax}} \) and \( J_{\text{max}} \), and is controlled similarly by LMA (i.e. by irradiance). If the latter conditions are not respected, then the model is rejected. \( J_{\text{max}} \) is still controlled by \( V_{\text{cmax}} \) and LMA (i.e. by irradiance). This model was not rejected either (Fisher’s C test: \( \chi^2 = 13.1, \text{ df} = 10, P = 0.220 \)).

Several other models have been checked and the general conclusions were: when a direct control of \( V_{\text{cmax}} \) by Nmass was hypothesized, then including a direct control of \( g_i \) or \( J_{\text{max}} \) by Nmass led to rejection of the models; a direct control of \( V_{\text{cmax}} \) on \( J_{\text{max}} \) had to be included in addition to a direct control of LMA on \( J_{\text{max}} \) otherwise the models were rejected. Interestingly, these models show that \( J_{\text{max}} \) is not solely determined by \( V_{\text{cmax}} \) but also and independently by LMA (i.e. by the irradiance gradient) too, invalidating the constancy of the ratio of \( J_{\text{max}} \) to \( V_{\text{cmax}} \). The absence of paths linking nitrogen content to any other variable but \( V_{\text{cmax}} \) indicates that nitrogen content determines the overall pool of proteins involved in carbon fixation but that its level does not influence the balance in relative allocation to the different photosynthetic functions. Besides, these models show that there are functional links between \( g_i \) and photosynthetic capacity, and confirm the functional links between light-saturated assimilation and mesophyll conductance evidenced by Terashima et al. (2001) in a structural–functional modelling approach.

**Seasonal variations**

The climate throughout the 2001 growth season had remained mild, and only a short and slight drought episode occurred during late August. This might have affected the outcome of the late August measurement campaign (Aug2). However, no indication of severe drought stress was noticed.

---

**Fig. 9.** Path models describing the functional relationships between LMA, Nmass, SPAD units (scaling to chlorophyll content on an area basis), \( V_{\text{cmax}}, J_{\text{max}}, \) and \( g_i \). See text for discussion.
when $A-C_i$ response curves were established: stomatal conductance and net assimilation (not shown) during the measurements were similar to those recorded during the other measurement campaigns. There are nevertheless some indications for mild stress effects during this campaign: mesophyll conductance of shade leaves was lower than during all other campaigns and relative nitrogen allocation to bioenergetics was at its lowest. However, if a stress actually occurred, it affected only understorey trees, which have a small if not negligible contribution to whole canopy carbon fixation. Moreover, the decline in relative allocation of N to bioenergetics might simply be the continuation of the overall decrease observed from June to early August.

Once maximum values were reached in mid-May, 3–4 weeks before the first campaign, LAI remained stable until late October, 2 weeks after the last campaign. Similarly LMA was stable and, therefore, the only driving variable that may have changed during the course of the season is leaf nitrogen: its total amount could have been affected by reallocations either between leaves in the canopy, or between leaves and storage organs. Another change could have affected the relative allocation of N to the different photosynthetic processes. Total nitrogen amount in the canopy, as estimated from the intercept of the $N_{mass}$ versus LMA relationship, varied slightly: it increased from June to mid-summer and then decreased until early October. The observed changes did not reveal major translocation processes. Furthermore, the nitrogen gradient (i.e. the slope of the $N_{mass}$ versus LMA relationship) remained stable throughout the growth period, indicating that there was no noticeable nitrogen redistribution between the upper and lower canopy.

The only noticeable changes affected the relative allocation of nitrogen to the different photosynthetic processes at an intra-leaf scale. Relative allocation of N to light capture was very stable: the slope of the relationship between LMA and the ratio SPAD$/N_{area}$ remained stable throughout the growing season and the small but significant seasonal trend in the intercept reversely followed that of $N_{mass}$. This latter fact indicates that the absolute amount of N invested into light capture remained stable throughout the season, as was also observed by Demarez et al. (1999).

Seasonal variations of $V_{cmax}$ resulted mainly in smaller gradients and higher variability during early summer than during late summer and autumn. This advocates for a slow acclimation to changes in the irradiance gradient (Frak et al., 2001) due to leaf emergence and expansion in spring. This is confirmed by the reversion of the gradient of relative N allocation to carboxylation observed between June and July and the following months. A general and important decrease of $V_{cmax}$ and of nitrogen allocation to carboxylation was observed during the last October campaign, while $N_{mass}$ decreases only weakly at that time. This suggests that autumn senescence was beginning and resulted in deactivation if not degradation of Rubisco.

Seasonal variations of the $J_{max}$ profiles were less marked: the gradients remained the same throughout the growth season, and the seasonal course of the scaling parameter was similar to that of $N_{mass}$ until the end of summer but not in October where $J_{max}$ increased whereas $N_{mass}$ decreased. This was confirmed by the evolution of the nitrogen allocation to bioenergetics (Fig. 7B). The increase of $J_{max}$ and, in parallel, of nitrogen allocation to bioenergetics in October remained difficult to interpret.

Some changes occurred throughout the growth season in $g_i$ gradients but these remained small compared with the large variability of these data, and no sound interpretation could be elicited except for the $g_i$ decrease in shade leaves in late August that was possibly attributed to the small drought stress occurring at that time, as discussed above.

Several studies reported a very visible seasonal time course of apparent $V_{cmax}$ in sun leaves of broad-leaved deciduous forests and claimed the importance of considering the seasonal trends of photosynthetic capacity when modelling ecosystem carbon fluxes. However, no universal trend could be drawn from such results. Some report a peak in early summer followed by a slow decrease until the beginning of senescence when the decrease accelerates (Wilson et al., 2000). Others evidenced a rapid increase in spring followed by a slow increase during summer before the autumnal, senescence-induced decrease (Muraoka and Koizumi, 2005) or a dome-like evolution peaking during mid-summer (Kosugi and Matsuo, 2006; Wang et al., 2008).

In the present study, there was no evidence of any time course of $V_{cmax}$ in the upper leaves (LMA >80 g m$^{-2}$) during summer except an increase between the June and July campaigns and a decrease in early October. Many factors might be involved in these differences, such as species behaviour, site and climate effects, and occurrence and intensity of drought events. Severe decreases in photosynthetic capacity were reported in longer lived sun leaves of evergreen species; they mainly occurred during the second growth season (Katahata et al., 2007). In fully annual broad-leaved species with rather short-lived leaves, this may not be the case This question still deserves further research.

Conclusions

This study confirmed the primary role of structural adjustment through the plastic response of LMA for the acclimation of leaves to the local irradiance in a canopy, whereas biochemical adjustments through relative nitrogen allocation remained marginal. A strong co-ordination was evidenced between $V_{cmax}$, $J_{max}$, and mesophyll conductance, and was only slightly modulated by local irradiance. The plastic relationships between these parameters deserves further attention. In the absence of severe drought stress during summer, photosynthetic capacity proved to be quite stable in summer after a leaf expansion phase and before autumn senescence. A steep vertical gradient of mesophyll conductance was evidenced and should be considered when using multi-layered model of carbon fluxes in forest ecosystems.

Acknowledgements

The help of Delphine Retzinger and Michèle Ridolfi during measurement campaigns is gratefully acknowledged, as well
as the contribution of Jacqueline Marchand in analysing carbon and nutrient contents. Funding for this study was provided by the European program Carboeuroflux (EVK2-1999-00229). Helpful discussions about g, with B Genty (CNRS-CEA, Cadarache, France) and CR Warren (University of Sydney), and about path analysis with B Shipley are gratefully acknowledged.

References


