Continuous Parameterization of Leaf Area Index and Phenological Phases Within Deciduous Forests Based on Temperature Measurements

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The Leaf-Area-Index (LAI) is commonly used to characterize the plant canopy and is a fundamental indication of plant vitality and photosynthetic activity. The forest health status is not only vital for economical reasons, but also has a significant impact on global carbon sequestration. The LAI has a highly dynamic character among deciduous forests and is prone to significant seasonal fluctuations. Accurate continuous LAI measurements do provide valuable information on growth characteristics, but they require considerable measurement effort. In this study, we tested a novel method that would allow for continuous low-effort LAI parameterizations. For our study we used temperature measurements from 2011 to 2019 obtained at two meteorological stations: Station one is an open land station, station two is located inside a forest stand characterized by European beech (measurements were undertaken as part of the ICP Forests program), both are located in Klausen Leopoldsdorf (Austria). We chose the difference in daily maximum temperature between the two sites for our LAI parametrization ($\text{LAI}_{\text{par}}$) since the forest canopy has a significant impact on local radiation conditions. We were able to identify phenological events such as leaf unfolding, the end of leaf growth, and the beginning and end of defoliation by examining at the average course of the year for $\text{LAI}_{\text{par}}$. The resultant $\text{LAI}_{\text{par}}$ values were compared to annual values derived from hemispheric photographs taken near the stand temperature sensor. For the years 2011–2017, we found a strong correlation of 0.93 between LAI measures and $\text{LAI}_{\text{par}}$, which dropped to 0.69 after adding the year 2018 and 0.32 after adding 2019. We further compared the phenological events obtained from $\text{LAI}_{\text{par}}$ to phenological observations. The impact of forests on their site climate, according to our findings, can be utilized to identify phenological and growth characteristics. The proposed method, however, is not a replacement for conventional LAI measurements.

Keywords: leaf area index, phenology, foliage, microclimate, forest climate, fagus sylvatica
1. INTRODUCTION

The forest canopy is an important interface between the terrestrial ecosystem and the atmosphere allowing for an exchange of energy, water, and carbon dioxide (Bonan, 1993). This interface may also be affected by climate change as previous studies have shown that the occurrence of phenological phases has shifted due to temperature changes (Chmielewski and Rötzer, 2001; Cleland et al., 2007). That being said the determination and quantification of phenological features and forest canopy extent are non-trivial and oftentimes elaborate:

1.1. Monitoring of Phenological Phases

Cleland et al. (2007) discussed limitations and opportunities of different methods to monitor plant phenology. The traditional method for detecting phenological phases relies on the utilization of observers (oftentimes volunteers) that register days that are associated with changes in phenological phases such as leaf unfolding during spring, bud formation, flowering, or leaf coloring and defoliation in late summer and fall. The phenological network provides conditions and thresholds to identify the various phases. One of the main problems arising when using data from phenological observations is the low spatial and temporal resolution. Phenological data can also be obtained using eddy-covariance flux measurements. As atmospheric carbon dioxide oscillates on an annual cycle, the amplitude and shift in this cycle can be used as an indicator of plant phenology. The collected data is however site-specific and in general, the setup for eddy-covariance monitoring is very expensive. Other ground and near-surface methods include digital camera sensing, spectral radiometers, and cameras carried by Unmanned Aerial Vehicles (Berra and Gaulton, 2021). The availability of reliable data is especially crucial for the evaluation and improvement of emerging, generally less complex technologies like phenology modeling and remote sensing. Phenology modeling allows not only for future estimates but also for the reconstruction of time series into the past where data is scarce. Limitations arise because the number of physiological processes described in the models is limited. Berra and Gaulton (2021) reviewed monitoring of forest phenology based on remote sensing which is oftentimes referred to as the Land-Surface-Phenology (LSP). Plant phenology shifts occur at a rate of 1 day per decade on average, however, the usual uncertainties from satellite LSP are much larger. This suggests that utilizing satellite data to track phenological changes still has considerable limitations.

1.2. Monitoring of Canopy Extent

The forest canopy extent can be described by numerous different parameters, with the most commonly used being the Leaf-Area-Index (LAI) which is defined as the projected area of leaves per unit of ground (Ross, 2012). LAI was shown to affect local temperature and vapor pressure deficit conditions, creating a characteristic forest microclimate (Von Arx et al., 2013). De Frenne et al. (2013) suggested that microclimatic buffering in temperature might help to decelerate a species turnover to the benefit of species from warmer regions (thermophilization) which is expected to occur due to macroclimate warming caused by climate change whilst mentioning that high closure in forest canopy might decrease the occurrence of light-demanding understory plants in forests. In a more recent approach, De Frenne et al. (2019) described the microclimatic buffering to be a potentially useful functioning to protect biodiversity. When deciding on a location and time for a field campaign, the LAI’s dynamical nature must be taken into account: The main driver of spatial LAI variations between stands and years has previously been identified to be forest management, as thinning and harvesting drastically alter stand structure. Le Dantec et al. (2000). Aside from the high spatial heterogeneity, LAI is changing dramatically on various time scales. Besides year-to-year differences in the extent and natural seasonal variations especially in deciduous forests disturbances due to climatic or biotic factors can also have a major influence. Measurements of LAI are classified into direct and indirect methods. Breda (2003), Jonckheere et al. (2004), and Weiss et al. (2004) reviewed advantages and disadvantages of different measurement techniques in detail. Direct methods vary in their sampling destructiveness. Destructive methods can involve the removal of leaves from one or more model trees. Obtained LAI values are then up-scaled under the assumption of homogeneity within the stand. For a less destructive direct measurement littertraps can be placed and evaluated. The main disadvantage of direct methods is that they are in general time-consuming and labor-intensive and therefore only partially applicable for detailed long time monitoring. However, they are still needed for the calibration and validation of indirect methods. Indirect methods can be divided into contact and non-contact measurements. One commonly used indirect contact measurement is the utilization of allometric relationships between LAI and other parameters such as the sapwood area at breast height. These relationships however are not universally applicable since they are dependent on stand characteristics (Mencuccini and Grace, 1995; Le Dantec et al., 2000). An important group of non-contact indirect measurement methods are Beer’s law-based optical methods such as hemispherical photographs. The setup for this method is usually a high-resolution digital camera equipped with a wide-angle fisheye lens mounted on a self-leveling system. The manually taken images are analyzed using software packages. Yan et al. (2019) described the three main issues of optical measurements to be the Leaf angle-distribution (LAD), clumping effects, and woody components. Moreover, Zheng and Moskal (2009) noted that a potential error source is the manual setting of exposure values if a camera is being used. Another issue when using photographs is that they only provide two-dimensional information. LiDAR measurements offer point cloud information, that could be used to retrieve three-dimensional data (Yan et al., 2019). It is distinguished between three different LiDAR setups: terrestrial laser scanner (TLS),
airborne laser scanner (ALS), and space-borne laser scanner (SLS). The latter of which is particularly relevant for generating LAI information on a global scale (Tang et al., 2014). The LAI can also be derived from satellite data utilizing vegetation reflectance properties. Typically, in that case, satellite-derived vegetation indices such as the normalized difference vegetation index (NDVI) (Rouse et al., 1974) or the enhanced vegetation index (EVI) (Huete et al., 2002) are being used to compute LAI estimates. Potithep et al. (2013) stated that seasonal changes are the most likely source of uncertainty in broadleaf forests and that it still requires in situ data of not just LAI but also vegetation indices to improve satellite-based LAI estimates. They further compared the determined LAI values to in situ measurements using NDVI and EVI data from MODIS images and discovered that EVI estimates produced better outcomes, particularly during the leaf growth phase. Non-green leaves cause mistakes in both VIs (from the saturation to the leaf fall period).

The objective of this study is to examine the modification of temperature inside a beech stand in comparison to open land conditions and analyze to which extent microclimatic differences can be used to quantify the status of the forest canopy. The presented method provides an LAI parametrization on a day-to-day basis which allows depicting seasonal changes. Furthermore, year-to-year differences in the occurrence and duration of phenological phases can be assessed. As stand data used in this study is collected as part of the ICP forests monitoring program similar time-series data are available in several European countries. Replicating the analysis might not only allow for detailed comparisons between different regions but moreover, results could be used as a reference when analyzing phenological phases determined using remote sensing technology. The novelty of this method is the generation of detailed phenological data on an operational basis.

2. MATERIALS AND METHODS

2.1. Site and Field Measurements

This study focuses on a beech stand (Fagus sylvatica L.) in Klausen-Leopoldsdorf (Austria, 48°07′16″N 16°02′52″E, 510 msl) during the years 2011–2019. This site is a core plot of the Level II Monitoring initiative (de Vries et al., 2003; Neumann and Kindermann, 2016). Level II monitoring is an initiative of the ICP Forests program which started in 1995. In Austria 16 intensive observation areas were established. Six of which were selected as core plots, where additional parameters are being measured. Among others, the core plots supply data on meteorological conditions inside the forest stand. We used daily maximum temperature ($T_{d}$) inside the forest stand which was measured in agreement with the programme standards using a temperature sensor placed 2 meters above ground at measurement intervals of 15 min.

As a reference for open land conditions we used daily maximum temperature ($T_{d}$) provided by the Austrian National Weather Service (ZAMG). The station is part of the national weather monitoring service (TAWES) and located in open land also in Klausen-Leopoldsdorf in proximity to the Level II site (48°03′6.84″N 16°00′2.159″E, 389msl). Measurements are recorded in an interval of 10 min.

An overview of both stations and the research area can be found in Figure 1, maximum temperature characteristics for both stations can be found in Table 1. It is important to note that the lapse rate between the two stations has not been taken into account since it has no significant influence on the results and findings of our study.

For the determination of LAI values we used hemispherical photographs which were also conducted as part of the ICP Forests program. The photographs were taken in the near vicinity of stand temperature measurements. To get more representative LAI estimations, photographs were taken at 16 different photographic points in a 10 × 10 meter grid. We used a Canon EOS 500 D camera fixed on a 1.3 m tripod which was leveled and aligned with the magnetic North. We used the programme WinSCANOPY 2009a with the recommended settings to analyze images. Photographs were obtained once a year between the 3 and 29. August. The obtained LAI ($LAI_{hem}$) was on average at 1.96 with an overall standard deviation of 0.33. The maximum $LAI_{hem}$ of 3.41 was found in 2012 where the standard deviation between the photographic points was with 0.41 also at maximum. The minimum value of 1.26 was found in 2012.

Phenological observations of leaf unfolding and leaf fall for Fagus sylvatica L. were also used for comparison. Since no data from fixed observation points was available we calculated
the median day of observation in an area stretching in West-
East direction from 14°45′57.9492″O to 17°9′38.7′′O and
stretching in North-South direction from 48°30′35.334″N to
46°59′12.0084″N. Observations were provided by the ZAMG and
obtained as part of the PhenoWatch programme (Templ et al.,
2018). The point in time when at least three parts of the observed
tree leaves have completely unfolded and in their final form (but
not size) was classified as leaf unfolding. Leaf fall was defined as
the point at which roughly half of the crown had defoliated.

2.2. Methodology
For this analysis Python version 3.7 was used (Van Rossum
and Drake, 2009) in addition with the packages numpy (Harris
et al., 2020), pandas (Wes McKinney, 2010; pandas development
team, 2020) and scipy (Virtanen et al., 2020) for data analysis
as well as matplotlib (Hunter, 2007) and seaborn (Waskom, 2021)
for visualization.

2.2.1. Foliage During the Course of the Year
Forest canopy closing affects stand radiation conditions.
As a result, the daily maximum temperature, which is
highly dependent on incoming radiation, changes. To quantify
this effect, we calculated the difference in daily maximum
temperature inside the stand \( T_\text{F} \) and at the reference site \( T_\text{R} \)
as \( \text{DMAX} \):

\[
\text{DMAX} = T_\text{F} - T_\text{R}
\]

To achieve an LAI parametrization that is consistent with the
expected foliage course in deciduous forests (low closure during
the winter, increased during the spring, maximum closure during
the summer, and decrease at the end of the growing season), we
implemented a sign convention and defined the parameterized
LAI (\( \text{LAI}_{\text{par}} \)) as:

\[
\text{LAI}_{\text{par}} = -\text{DMAX}
\]

In order to study the average seasonal variation of \( \text{LAI}_{\text{par}} \)
and compare the obtained curve with the phenological observations
of leaf unfolding and leaf fall, the median of \( \text{LAI}_{\text{par}} \) was calculated
for each individual day of the year over all available years. In
addition, Gompertz function was used to further explore foliage
features, particularly during the growth and defoliation phases:

\[
y = A \exp(-\exp(\beta - \kappa t))
\]

\( y, \text{LAI}_{\text{par}} \) normalized between -1 and 1; t, time (expressed as day
of the year) normalized between -1 and 1; A, upper asymptote;
\( \beta \), x-axis placement parameter; \( \kappa \), rate of change of shape (Rossi
et al., 2003).] To determine the parameters \( A, \beta \) and \( \kappa \) the curve
fit function implemented in the scipy package was used which
fits a function based on non-linear least squares. To determine
the inflection point \( t_\text{ip} \) the second derivative has to be set equal
to zero, leading to:

\[
y'' = y'\kappa \exp(\beta - \kappa t) - 1 = 0
\]

and thus it can be found where \( t_\text{ip} = \beta / \kappa \) (Rossi et al., 2006).
This analysis was performed for two growth phases individually:

leaf growth (timespan day 1–day 240) and defoliation (timespan
day 160–day 366). The timespans were chosen so that each
one contains the transition from one constant state to another
constant state which is essential for the Gompertz Fit (for the
leaf growth phase from no foliage to maximum foliage, for the
defoliation phase from maximum foliage to no foliage).

2.2.2. Modulating Effect of Forest Canopy on Daily
Maximum Temperatures
In order to further investigate diminished heating inside the
forest stand we normed DMAX using \( T_R \):

\[
\text{DMAX}_\text{N} = \frac{\text{DMAX}}{T_R}
\]

As described above, we calculated and plotted the median course
of the year. For our analysis, we mainly focused on \( \text{DMAX}_\text{N} \)
during summertime, since the moderating effect of forest cover
is especially of interest during the warm season.

2.2.3. Comparison to \( \text{LAI}_{\text{hem}} \)
Contrary to the \( \text{LAI}_{\text{par}} \) values that are available once a day,
\( \text{LAI}_{\text{hem}} \) values are obtained only once a year. In order to extract
one \( \text{LAI}_{\text{par}} \) value for annual comparison, we decided on using
the \( \text{LAI}_{\text{par}} \) values obtained at the exact days that the hemispheric
photographs were taken. This was preferred over using weekly
or monthly averages, as it is important to compare \( \text{LAI}_{\text{par}} \) at
similar radiation conditions (Preferably no rain and fair weather
conditions) which should be the case on the days the hemispheric
photographs were taken. To allow a reasonable comparison
between \( \text{LAI}_{\text{par}} \) (that is essentially a temperature difference) and
\( \text{LAI}_{\text{hem}} \) we standardized both annual timeseries using:

\[
x_{\text{standardized}} = \frac{x - \text{mean}}{\text{std}}
\]

Where in this case \( x \) represents the individual measurement,
\( \text{mean} \) the mean value and \( \text{std} \) the standard deviation of \( x \)
calculated from the total time series. We observed an extreme
drop in performance when we included data from the year 2019
where an outlier in \( \text{LAI}_{\text{par}} \) occurred. Hence, we decided to include
plots and statistics for the time spans 2011–2017, 2011–2018,
and 2011–2019 for comparisons.

3. RESULTS

3.1. Foliage During the Course of the Year
The median course of the year for \( \text{DMAX} \) can be found in the
Supplementary Material (Supplementary Figure S1). The plot
was sectioned into two different foliage phases: The “Maximum
foliage phase” ranging from 8.June to 27.August (160–day 240)
and the “No foliage phase” ranging from the 20.November to
the 4.April (325–day 95). This process is relatively subjective and
only gives approximate dates since it was based solely on the form
of the curve. The median value of \( \text{DMAX} \) is at \(-2.55°C\) with a
standard deviation of \(0.89°C\). The median during the no foliage
phase in late autumn, winter and early spring is at \(-1.92°C\) with a
standard deviation of \(0.27°C\). At the maximum foliage phase the
median is at $-3.86^\circ C$ with a standard deviation of $0.33^\circ C$. The median course of the year for $LAI_{par}$ can be found in Figure 2. Due to the definition of this parameter standard deviation values are the same as for $DMAX$, median values differ only in sign. The fit determined by applying the gompertz function to $LAI_{par}$ values including measurement values can be seen in Figure 3, the function parameters can be found in Table 2. Characteristic days namely the day of maximum leaf growth (day 135) and maximum defoliation (day 289) were marked by vertical lines in the plot. $LAI_{par}$ values and characteristic days for individual years were calculated and plotted. They can be found in the Supplementary Figure S2, Table S1 (for the years 2011–2015) respectively Supplementary Figure S3 (for the years 2016–2019).

3.2. Moderating Effect of Forest Canopy on Daily Maximum Temperatures

During the measurement period the median value for $DMAX_N$ is at $-0.17$ with a variance of $0.42$. During the no foliage phase a median of $-0.35$ with a variance of $1.05$ was obtained. It is important to note that high variation during this phase can partly be caused by maximum temperature approaching $0^\circ C$ on some days. During the maximum foliage phase the median of $DMAX_N$ remained relatively constant at $-0.15$ (with a variance of $0.0002$). The constancy during the maximum foliage phase is especially of interest, since formula 5 can be rearranged to:

$$T_F = T_R + (DMAX_N \times T_R) \quad (7)$$

Formula 7 emphasizes that $DMAX_N$ describes the effective moderating effect the forest canopy has on forest maximum temperatures. When determining $DMAX_N$ for the maximum foliage phase of individual years, minimal variation was found ranging from $-0.17$ (2016) to $-0.12$ (2017). Detailed averages and variances for the maximum foliage phase in each individual year can be found in Table 3. Because the value of $DMAX_N$ during the maximum foliage phase showed minimal variation
TABLE 3 | DMAXh median values and variances for individual years during the maximum foliage phase.

| YEAR | Median | Variance |
|------|--------|----------|
| 2011 | -0.14  | 0.045    |
| 2012 | -0.15  | 0.026    |
| 2013 | -0.15  | 0.025    |
| 2014 | -0.15  | 0.030    |
| 2015 | -0.15  | 0.014    |
| 2016 | -0.17  | 0.051    |
| 2017 | -0.12  | 0.099    |
| 2018 | -0.14  | 0.081    |
| 2019 | -0.14  | 0.046    |

FIGURE 4 | Relationship between the observed $T_F$ and the approximation using DMAXh and $T_R$.

3.3. Comparison to LAI_{hem}
A comparison between standardized LAI_{hem} and LAI_{par} values can be found in Figure 5. Boxplots were used to depict the variation of LAI_{hem} for the different photographic points. The comparison of annual values can be found in Table 4. To further investigate the influence of the year 2019 three scatter plots were created, showing the relationship between annual average LAI_{hem} and LAI_{par} for the timespans 2011–2017, 2011–2018, and 2011–2019, which can be found in Figure 6. Corresponding regression parameters as well as calculated Spearman correlations can be found in Table 5. Spearman correlation between LAI_{hem} and LAI_{par} was very strong obtaining a value of 0.93 with a $p$-value of 0.006 during the time span 2011–2017. A moderate correlation of 0.69 with a $p$-value of 0.021 was found during 2011–2018. During 2011–2019 correlation became negligible obtaining a value of 0.32 with a $p$-value of 0.757.

4. DISCUSSION
4.1. Foliage During the Course of the Year
By analyzing the median course of the year for LAI_{par} (Figure 2) we were able to identify two constant foliage phases: The “Maximum foliage phase” during which the forest canopy is fully developed and no defoliation has yet taken place and the “No foliage phase” for which the constant low temperature difference can be interpreted as the blocking of solar radiation by the branches when defoliation is completed. Comparing the timing of the end of the no foliage phase with observed leaf unfolding showed a delay, with the no foliage phase ending on day 95 and the observed leaf unfolding being marked as day 108. However, this was expected as leaf unfolding was defined within the PhenoWatch program as the day where at least three parts of the crown leaves have already unfolded and have their final shapes and LAI_{par} will potentially also react to smaller disturbances such as bud formation and unrolling. For leaf fall a similar problem arises. As the observed leaf fall is at day 280 whereas the no foliage phase starts 45 days later. Per definition, this observation occurs when at least 50% of the crown is defoliated. However, the remaining crown is still blocking radiation and therefore affects DMAX, respectively, LAI_{par}. This could be the reason why leaf fall was observed before the beginning of the no foliage phase. The $R^2$ for the Gompertz function fit was at 0.68 during the leaf growth phase and at 0.63 for the defoliation phase. We compared the occurrence of the main phenological events as well as duration of phenological phases determined using the presented method to the results of previous studies on Fagus sylvatica L. foliage throughout the year:

- Global radiation profiles used by Wang et al. (2004), obtained in a stand in Hesse, France throughout the years 1996–2001.
- Moderate Resolution Imaging Spectrometer (MODIS) used by Wang et al. (2005) in a stand in Hesse, France throughout the years 2001–2003.
- Hemispherical photographs and visual assessments of bud development used by Bequet et al. (2011) in 10 European beech stands distributed in Flanders Belgium in 2008 for a detailed analysis of leaf development phases.

A detailed comparison of characteristic days and duration of different foliage phases obtained via the four methods can be found in Table 6. Since the data sets come from very different climatic regions and from different years, this comparison is only used to check if characteristic days determined using the presented method are plausible. The determined day for the end of the no foliage phase using the LAI_{par} method is prior to the day determined using hemispherical photographs or radiation profiles and subsequent to the day determined using MODIS. Wang et al. (2005) argued that an early onset date for MODIS is found because it detects the onset of greens...
rather than a species specific event. It can not be ruled out that greening has an effect on the local temperature conditions and thereby also on LAI_par however the magnitude of this influence is unclear. The start of the maximum foliage phase was late in comparison to the day determined using radiation profiles and early in comparison to the day obtained using hemispheric photographs. The start of the no foliage phase was approximately the same as determined by MODIS. Considering the high variation of the determined phenological events between the different methods, the presented LAI_par method leads to plausible results. Bigger differences occurred when comparing the duration of specific phases. Where duration of the leaf growth phase was remarkably longer than duration obtained using the radiation model and duration of the maximum foliage phase remarkably shorter. Overall our findings suggest that LAI_par can be used to determine when considerable microclimatic changes are arising from foliage change within a deciduous forest.

4.2. Moderating Effect of Forest Canopy on Daily Maximum Temperatures

We found the relationship between the observed and approximated $T_F$ to be highly significant and concluded that $DMAX_N$ can provide insight into stand-specific microclimate conditions and can be used to highlight the recreational function of forest land during summer. As an example, a hypothetical temperature prediction of $35^\circ C$ for open land corresponds to a lower maximum temperature by $5,25^\circ C$ inside the stand. For individual years this reduction would range between $4.2^\circ C$ (2017) and $5.95^\circ C$ (2016).

Diminished daytime heating inside forest stands is of particular interest in regions where climate predictions suggest that the frequency of heat days and heatwaves is increasing. $DMAX_N$ is presumed to vary depending on stand characteristics such as forest type, dominant species, altitude, and slope, which are known to have a considerable impact on temperature moderation inside forest stands (Renaud and Rebetez, 2009; Ferrez et al., 2011; Renaud et al., 2011).

4.3. Comparison to LAI_hem

Correlations were very strong during the time span 2011–2017. During the timespan 2011–2018 correlation was moderate. We identified a few possible reasons for the decline in correlation when including the year 2019 to our analysis:

- Inhomogeneities in the radiation conditions. Since we evaluated the temperature difference at the exact day the hemispheric photographs were taken it is possible that radiation conditions varied between the years. Aggregating over more days by building averages for 1 week, month or within the maximum foliage phase however does not necessary lead to a more robust parametrization since radiation conditions can differ drastically.
- Spatial inhomogeneities in comparison to the area where LAI_hem is measured. Temperature and LAI_hem are both point measurements, however through temperature mixing the LAI_par contains not only information on the forest canopy status directly above the measurements but moreover within a nearby region (depending on local conditions such as wind).
- Malfunctioning of the stand temperature sensor. Since measurements malfunctioned after the 22nd of September in 2019 issues with the temperature sensor inside the stand could cause the extreme low value of LAI_par.

Since LAI_hem measurements are only available annually and the period of the analysis is limited to seven years, caution should be
excercised when interpreting the results of this section. On the one hand, the correlations during 2011–2017 and 2011–2018 seem to be promising, on the other hand, the drop in correlation when adding data from 2019 cannot be disregarded. To assess how well LAI_{par} is suitable as an LAI parameterization, the collection of more LAI data is needed, preferably also during different foliage phases. Our results suggest that LAI_{par} cannot replace conventional LAI methods but can provide interesting insights on stand foliage and microclimate conditions.

**DATA AVAILABILITY STATEMENT**

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

**AUTHOR CONTRIBUTIONS**

AZ and SS designed the study. AZ analyzed the data. KG provided stand data to the study. GS provided data of phenological observations to the study. All authors contributed to the article and approved the submitted version.

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**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/ffgc.2021.768085/full#supplementary-material
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