Can a satellite-derived estimate of the fraction of PAR absorbed by chlorophyll (FAPAR_{chl}) improve predictions of light-use efficiency and ecosystem photosynthesis for a boreal aspen forest?

Qingyuan Zhang, Elizabeth. M. Middleton, Hank A. Margolis, Guillaume G. Drolet, Alan A. Barr and T. Andy Black

Gross primary production (GPP) is a key terrestrial ecophysiological process that links atmospheric composition and vegetation processes. Study of GPP is important to global carbon cycles and global warming. One of the most important of these processes, plant photosynthesis, requires solar radiation in the $0.4-0.7 \mu m$ range (also known as photosynthetically active radiation or PAR), water, carbon dioxide (CO₂), and nutrients. A vegetation canopy is composed primarily of photosynthetically active vegetation (PAV) and non-photosynthetic vegetation (NPV; e.g., senescent foliage, branches and stems). A green leaf is composed of chlorophyll and various proportions of nonphotosynthetic components (e.g., other pigments in the leaf, primary/secondary/ tertiary veins, and cell walls). The fraction of PAR absorbed by whole vegetation canopy (FAPAR_{canopy}) has been widely used in satellite-based Production Efficiency Models to estimate GPP (as a product of FAPAR_{canopy} x PAR x LUE_{canopy}, where LUE_{canopy} is light use efficiency at canopy level). However, only the PAR absorbed by chlorophyll (a product of FAPAR_{chl} x PAR) is used for photosynthesis. Therefore, remote sensing driven biogeochemical models that use FAPAR_{chl} in estimating GPP (as a product of FAPAR_{chl} x PAR x LUE_{chl}) are more likely to be consistent with plant photosynthesis processes.

Our paper has been designed to test which group ($[FAPAR_{canopy}, LUE_{canopy}]$ vs. [FAPAR_{chl}, LUE_{chl}]) is more consistent with plant photosynthesis processes.

Using a coupled canopy-leaf radiative transfer model, we have estimated FAPAR_{chl} and FAPAR_{canopy} for the Southern Old Aspen forest (SOA) in Canada for 2001-2005 with MODIS images. The tower fluxes over the SOA site provide real time photosynthesis of the forest. The scientists of the SOA site offer the measurements of photosynthesis and their flux tower based LUE (LUE_{tower}). Our results showed that LUE_{chl} matched well with LUE_{tower} both at magnitude and at phase while LUE_{canopy} did not. Using FAPAR_{canopy} to estimate absorbed PAR for photosynthesis will significantly overestimate. One can't get good estimate of GPP phenology if using the group of FAPAR_{canopy} and LUE_{canopy} to estimate GPP. Using FAPAR_{chl} and LUE_{chl} to estimate GPP will get more consistent results and will help the study of global carbon cycles and global warming.

Can a satellite-derived estimate of the fraction of PAR absorbed by chlorophyll (FAPAR_{chl})

improve predictions of light-use efficiency and ecosystem photosynthesis for a boreal aspen

forest?

Qingyuan Zhang^{a,b}, Elizabeth M. Middleton^b, Hank A. Margolis^c, Guillaume G. Drolet^c,

Alan A. Barr^d and T. Andrew Black^e

^aGoddard Earth Sciences & Technology Center, University of Maryland, Baltimore County, Baltimore, Maryland 21228, USA

^bBiospheric Sciences Branch, Code 614.4, NASA/Goddard Space Flight Center, Greenbelt, Maryland 20771, USA

[°]Centre d'Étude de la Forêt (CEF), Faculté de Foresterie et de Géomatique, Pavillon Abitibi-Price, Université Laval, Québec City, Québec, Canada G1K 7P4

^dAtmospheric Sciences and Technology Directorate, Meteorological Service of Canada, Saskatoon, Saskatchewan S7N 3H5 Canada

^eBiometeorology and Soil Physics Group, 121-2357 Main Mall, University of British Columbia, Vancouver, BC, Canada V6T1Z4

Corresponding author:

Qingyuan Zhang

Address:

Building 33, Room G321 Biospheric Sciences Branch, Code 614.4, NASA/Goddard Space Flight Center, Greenbelt, Maryland 20771, USA

Tel: 301-614-6672 Email: qyz72@yahoo.com

2 Abstract

We used daily MODerate resolution Imaging Spectroradiometer (MODIS) imagery obtained 3 over a five-year period to analyze the seasonal and inter-annual variability of the fraction of 4 absorbed photosynthetically active radiation (FAPAR) and photosynthetic light use efficiency 5 (LUE) for the Southern Old Aspen (SOA) flux tower site located near the southern limit of the 6 boreal forest in Saskatchewan, Canada. To obtain the spectral characteristics of a standardized 7 land area to compare with tower measurements, we scaled up the nominal 500 m MODIS 8 products to a 2.5 km x 2.5 km area (5x5 MODIS 500 m grid cells). We then used the scaled-up 9 MODIS products in a coupled canopy-leaf radiative transfer model, PROSAIL-2, to estimate the 10 fraction of absorbed photosynthetically active radiation (APAR) by the part of the canopy 11 dominated by chlorophyll (FAPAR_{chl}) versus that by the whole canopy (FAPAR_{canopy}). Using 12 the additional information provided by flux tower-based measurements of gross ecosystem 13 production (GEP) and incident PAR, we determined 90-minute averages for APAR and LUE 14 (slope of GEP:APAR) for both the physiologically active foliage (APAR_{chl}, LUE_{chl}) and for the 15 entire canopy (APAR_{canopy}, LUE_{canopy}). 16

The flux tower measurements of GEP were strongly related to the MODIS-derived 17 estimates of APAR_{chl} ($r^2 = 0.78$) but only weakly related to APAR_{canopy} ($r^2 = 0.33$). Gross LUE 18 between 2001 and 2005 for LUE_{chl} was 0.0241 µmol C µmol⁻¹ PPFD whereas LUE_{canopy} was 19 36% lower. Time series of the 5-year normalized difference vegetation index (NDVI) were used 20 to estimate the average length of the core growing season as days of year 152 - 259. Inter-annual 21 variability in the core growing season LUE_{chl} (µmol C µmol⁻¹ PPFD) ranged from 0.0225 in 22 2003 to 0.0310 in 2004. The five-year time series of LUE_{chl} corresponded well with both the 23 seasonal phase and amplitude of LUE from the tower measurements but this was not the case for 24

I	LUE_{canopy} . We conclude that LUE_{chl} derived from MODIS observations could provide a more
2	physiologically realistic parameter than the more commonly used LUE_{catopy} as an input to large-
3	scale photosynthesis models.
4	
5	Key Words: MODIS; aspen; chlorophyll; ecosystem flux; gross primary production; gross
6	ecosystem production; light use efficiency (LUE); LUE _{chl} ; FAPAR _{chl} ; NDVI; LUE _{tower}
7	
8	
9	
10	
11	
12	
13	
14	
15	
16	
17	
18	

1 1. Introduction

2 1.1 Background – Using Light Use Efficiency to Estimate Ecosystem Photosynthesis

Realistic models of plant canopy photosynthesis are necessary for obtaining accurate estimates of 3 the carbon cycle for use in land surface models (LSMs) and atmospheric general circulation 4 models (GCMs) (Sellers et al. 1996a, 1996b). In vegetative canopies, photosynthetically active 5 radiation (PAR) is absorbed from sunlight by photosynthetic pigments, primarily chlorophyll a 6 and its accessory pigments (chlorophyll b, carotenoids). When ecosystem photosynthesis is 7 calculated with a process model, it is referred to as Gross Primary Production (GPP). When it is 8 calculated from flux tower data, it is referred to as Gross Ecosystem Production, designated here 9 as GEP_{tower}. 10

Plant production efficiency models (PEMs) have been developed to estimate GPP at 11 canopy, landscape, regional and global scales, utilizing optical remote sensing to provide the 12 fraction of absorbed PAR (FAPAR). Examples include GLO-PEM (Prince et al. 1995, 2000; 13 Prince and Goward 1995, 1996), TURC (Ruimy et al. 1994, 1996a, 1996b), 3-PG (Landsberg 14 and Waring 1997; Law et al. 2000) and PSN (Running et al. 1994, 1999a, 1999b, 2000, 2004). 15 This latter model is a satellite-based global photosynthesis product derived from the MODerate 16 resolution Imaging Spectroradiometer (MODIS) on the Terra and Aqua platforms. 17 All of these models estimate GPP as the product of three terms: the light use efficiency of 18 the canopy (LUE_{canopy}), which is a measure of the PAR conversion efficiency into 19 photosynthetically fixed CO₂; the FAPAR of the canopy (FAPAR_{canopy}), which is estimated 20 using radiative transfer models and remote sensing data or using empirical relationship between 21 FAPAR_{canopy} and the normalized difference vegetation index (NDVI, Tucker 1979); and the 22 incident PAR where: 23

$$GPP = LUE_{canopy} * FAPAR_{canopy} * PAR.$$
(1)

Consequently, accurate estimates of FAPAR and LUE for ecosystems are essential for obtaining
 accurate GPP.

The LUE concept was initially developed for agricultural crops at harvest to determine 4 the conversion efficiency of available light into biomass (g C dry mass) over a full growing 5 season and is typically expressed in units such as g C MJ⁻¹ PAR (Monteith 1972, 1977). This 6 seasonal crop-level LUE represents a direct measure of the average conversion efficiency of all 7 above ground plant material (Gower et al. 1999), which is dominated by foliage for agricultural 8 crops. Eddy covariance flux towers have the capability to provide near-continuous 9 measurements of GEP - denoted as GEPtower, and absorbed PAR - denoted as APARtower (see 10 section 3.2.3 for more details, also see Krishnan et al., 2006), for an entire ecosystem for time 11 12 periods as short as 30 minutes. Consequently, these instrumented flux towers also provide nearcontinuous measurements of LUE, denoted as LUE_{tower}, over these same time periods as: 13

14
$$LUE_{tower} = \frac{GEP_{tower}}{FAPAR_{tower}} * PAR_{tower} = \frac{GEP_{tower}}{APAR_{tower}}.$$
 (2)

The LUE_{tower} is typically expressed as μ mol CO₂ μ mol⁻¹ PAR or μ mol C μ mol⁻¹ PPFD, where 15 PPFD is the photosynthetic photon flux density. For these tower-based calculations, incident 16 PAR is measured directly by radiometers attached to the flux tower and the FAPAR estimate is 17 based both on detailed canopy structural measurements and on the radiometer measurements 18 (Barr et al. 2007). An underlying assumption supporting the LUE retrieval through the MODIS 19 modeling approach is that the LUE_{canopy} used in the models is a good approximation of LUE_{tower}, 20 at least when the measurement footprint of the tower is roughly equivalent to the area of the 21 remote sensing pixel. Apparent ecosystem LUE obtained from flux tower measurements (i.e., 22

1	LUE _{tower}) directly describes the integrated response of the whole ecosystem to the prevailing
2	environmental conditions, as do remotely acquired spectral snapshots although these latter are
3	limited to specific acquisition times and viewing configurations.
4	On a canopy or ecosystem scale, GEP and APAR are typically linearly related (e.g.,
5	Waring et al., 1995), so that LUE can be determined from the slope of this relationship. This
6	apparent linearity results from multiple scattering within the canopy, which involves 3-D
7	contributions from foliage of multiple species and illumination conditions, as well as non-
8	photosynthetic material (e.g., limbs, trunks, cones, litter). On the other hand, comparable light
9	response curves for individual leaves of selected species yield non-linear responses for which the
10	initial slope of the linear portion of the curve describes the quantum efficiency (Mohr et al.,
11	1995). The quantum efficiency of individual leaves can also serve as an input to carbon cycle
12	models but a means of scaling it to the canopy level is still required.
12 13	models but a means of scaling it to the canopy level is still required. A common modeling approach is to set a maximum LUE for optimal environmental
13	A common modeling approach is to set a maximum LUE for optimal environmental
13 14	A common modeling approach is to set a maximum LUE for optimal environmental conditions (i.e., unstressed vegetation) and to simulate ecosystem responses when unfavorable
13 14 15	A common modeling approach is to set a maximum LUE for optimal environmental conditions (i.e., unstressed vegetation) and to simulate ecosystem responses when unfavorable environmental conditions occur (e.g., limitations of temperature, humidity, soil moisture, etc.)
13 14 15 16	A common modeling approach is to set a maximum LUE for optimal environmental conditions (i.e., unstressed vegetation) and to simulate ecosystem responses when unfavorable environmental conditions occur (e.g., limitations of temperature, humidity, soil moisture, etc.) through down-regulation of the maximum LUE to achieve an apparent LUE (Medlyn, 1998).
13 14 15 16 17	A common modeling approach is to set a maximum LUE for optimal environmental conditions (i.e., unstressed vegetation) and to simulate ecosystem responses when unfavorable environmental conditions occur (e.g., limitations of temperature, humidity, soil moisture, etc.) through down-regulation of the maximum LUE to achieve an apparent LUE (Medlyn, 1998). The MODIS GPP product, an output of the PSN model, has been compared with
13 14 15 16 17 18	A common modeling approach is to set a maximum LUE for optimal environmental conditions (i.e., unstressed vegetation) and to simulate ecosystem responses when unfavorable environmental conditions occur (e.g., limitations of temperature, humidity, soil moisture, etc.) through down-regulation of the maximum LUE to achieve an apparent LUE (Medlyn, 1998). The MODIS GPP product, an output of the PSN model, has been compared with measurements made at flux towers by several research groups. For instance, Turner and
13 14 15 16 17 18 19	A common modeling approach is to set a maximum LUE for optimal environmental conditions (i.e., unstressed vegetation) and to simulate ecosystem responses when unfavorable environmental conditions occur (e.g., limitations of temperature, humidity, soil moisture, etc.) through down-regulation of the maximum LUE to achieve an apparent LUE (Medlyn, 1998). The MODIS GPP product, an output of the PSN model, has been compared with measurements made at flux towers by several research groups. For instance, Turner and colleagues (Turner <i>et al.</i> 2003, 2004, 2006) found that the annual MODIS GPP totals calculated

(GEP_{tower}) suggesting that a more physiologically realistic method of estimating GPP could be
 useful.

3

4

1.2 Chlorophyll-based LUE (LUE_{chl})

5 Even though maximum leaf LUE can be strongly influenced by leaf chlorophyll concentration (e.g., Waring et al. 1995), it is less clear how canopy chlorophyll concentration 6 7 might influence apparent LUE at the ecosystem scale. Laboratory studies (Yoder and Waring 8 1994) have shown that variation in canopy total chlorophyll content of miniature Douglas fir canopies was significantly correlated with their photosynthesis, although the correlation was 9 higher for canopies exposed to full sun. Several other studies have shown a relationship between 10 leaf or canopy nitrogen concentration and light use efficiency at the ecosystem scale (Kergoat et 11 12 al. 2008, Ollinger et al. 2008), However, we believe that remote sensing techniques that evaluate chlorophyll rather than nitrogen could have even greater potential for estimating ecosystem light 13 use efficiency and GPP. 14

From a biochemical perspective, only the PAR absorbed by photosynthetic pigments 15 (designated as APAR_{chl}) enables photosynthetic processes, whereas the PAR absorbed by non-16 17 photosynthetic components such as boles, branches, stems, and litter is not used for CO₂ fixation. We designate chlorophyll-based FAPAR here as FAPAR_{chl}. By definition, APAR_{canopy} (the 18 product of FAPAR_{canopy} and PAR) is greater than APAR_{chl} (the product of FAPAR_{chl} and PAR). 19 For linking to remote sensing applications, estimates of APAR_{chl} should provide more realistic 20 21 GEP_{tower} and LUE_{tower} values than similar estimates using APAR_{canopy}. We define LUE based on APAR_{chl} versus APAR_{canopy} as follows: 22

23

1
$$LUE_{chl} = \frac{GEP_{tower}}{FAPAR_{chl} * PAR} = \frac{GEP_{tower}}{APAR_{chl}}.$$
 (3)

2
$$LUE_{canopy} = \frac{GEP_{tower}}{FAPAR_{canopy} * PAR} = \frac{GEP_{tower}}{APAR_{canopy}}.$$
 (4)

In earlier studies (Zhang *et al.*, 2005, 2006), an approach to estimate FAPAR_{chl} was
proposed using daily MODIS data. Since then, we have refined our algorithm to retrieve
FAPAR_{chl} from MODIS imagery using the modified PROSPECT-SAIL2 model, PROSAIL-2
(Zhang *et al.* 2005, 2006). The new version of this algorithm provides a statistical distribution of
likely FAPAR_{chl} values for each cloud-free MODIS observation.

In this article, we combine five years of flux, meteorological, and remote sensing data 8 from a boreal aspen flux site to attain the following four objectives: (1) to present a method for 9 estimating FAPAR_{chl} and FAPAR_{canopy} using single-date, scaled-up MODIS observations; (2) to 10 apply the FAPAR_{chl} and FAPAR_{canopy} algorithms to MODIS data acquired for 2001-2005 over 11 this aspen flux site in Saskatchewan; (3) to link our estimates of MODIS FAPAR_{chl} and 12 FAPAR_{canopy} to the tower-based observations of PAR and GEP so as to derive LUE on both a 13 unit cholorophyll area basis (LUE_{chl}, Eq.3 above) and for the whole canopy (LUE_{canopy}, Eq.4 14 above); and (4) to compare our LUE_{chl}, LUE_{canopy} and tower-based LUE estimates (i.e., LUE_{tower}) 15 to see if the LUE_{chl} could provide a more physiologically realistic input to land surface process 16 models. For this latter objective, we test the hypotheses that: (i) $LUE_{canopy} = LUE_{chl}$; (ii) LUE_{chl} 17 = LUE_{tower} ; and (iii) LUE_{canopy} = LUE_{tower} . 18

1 **2.** Data and site descriptions

2 2.1 Southern Old Aspen

The Southern Old Aspen forest (SOA) was established in 1919 after a forest fire in Prince 3 Albert National Park at the southern edge of the Canadian boreal forest (Barr et al. 2007). The 4 eddy flux tower site (53.7°N, 106.2°W, 600 m elevation) is located ~50 km northwest of Prince 5 6 Albert, Saskatchewan. SOA originated as part of the BOReal Ecosystem Atmosphere Study (BOREAS), and has continued operations under the Boreal Ecosystem Research and Monitoring 7 Sites (BERMS) project and the Canadian Carbon Program, formerly the Fluxnet-Canada 8 Research Network (FCRN). The vegetation around the tower site is primarily deciduous forest 9 10 dominated 90% by aspen (Populus tremuloides Michx.) with 10% balsam poplar (Populus balsamifera L.) and a dense understory approximately 2 m tall of hazelnut (Corvlus cornuta 11 Marsh) interspersed with green alder (Alnus crispa Pursch). The aspen stand extends for at least 12 3 km in all directions from the tower. The soil is mainly an orthic grey luvisol with a ~9 cm 13 surface organic layer. The terrain is basically level. Mid-growing season leaf area index (LAI) 14 under the tower varied from ~3.5 to 5.5. The climate is warm in summer (av. 17.5°C in July) and 15 cold in winter (av. -19.1°C in January). Average annual precipitation was 412 mm during 2001-16 2005, whereas the long-term 1951-1980 average annual precipitation was higher at 484 mm 17 18 (Griffis et al. 2003). A detailed analysis of the inter-annual variability of climatic factors at this site is presented in Barr et al. (2007). 19

20

21 2.2 GEP Data

GEP at SOA was estimated as the sum of Net Ecosystem Production (NEP) and
 ecosystem respiration. Respiration was modeled from night time and cold season NEP using soil

1	temperature. Details on CO ₂ flux measurement methodology and obtaining estimates of GEP
2	including gap filling can be found in Krishnan et al. (2006) and Barr et al. (2007).
3	
4	2.3 Daily MODIS data
5	Four MODIS daily products (v004) were used in this study: [1] surface reflectance
6	(MOD09GHK and MYD09GHK); [2] observation viewing and illumination geometry
7	(MODMGGAD and MYDMGGAD); [3] observation pointer information (MODPTHKM and
8	MYDPTHKM); and [4] reflectance data quality descriptors (MOD09GST and MYD09GST).
9	The MODIS imagery is nominally acquired with 500 m x 500 m spatial resolution at nadir, and
10	>500 m spatial resolution for off-nadir views. The MODIS daily land surface reflectance
11	product provides seven (of 36) spectral bands: red (620-670 nm, band 1), blue (459-479 nm,
12	band 3), green (545–565 nm, band 4), near infrared (NIR ₁ , 841–875 nm, band 2; NIR ₂ , 1230–
13	1250 nm, band 5), and short-wave infrared (SWIR ₁ , 1628–1652 nm, band 6; SWIR ₂ , 2105–2155
14	nm, band 7).
15	The MODIS daily observation viewing geometry product provides viewing and
16	illumination geometry information (view zenith angle, VZA; view azimuth angle; sun zenith
17	angle; and sun azimuth angle) at a nominal 1-km scale. The MODIS daily observation pointer
18	product provides a reference, at a nominal 500 m scale, linking observations that intersect each
19	pixel in the daily surface reflectance product to those given in the daily observation viewing
20	geometry product. The MODIS daily reflectance data quality product provides summary quality
21	information about MODIS daily surface reflectance conditions, including clouds, cloud shadow,
22	land and water designations, aerosols, fire, snow, ice and bidirectional reflectance distribution
23	function (BRDF) corrections, etc. All the MODIS data products are freely available at USGS

Earth Observing System Data Gateway (<u>http://edcimswww.cr.usgs.gov/pub/imswelcome/</u>), and are delivered to users in a tile fashion, where each tile covers an area of 10° (latitude) by 10° (longitude). The software developed by the MODIS land team (MODLAND Tile Calculator <u>http://modland.nascom.nasa.gov/cgi-bin/developer/tilemap.cgi</u>) was utilized to determine the location of the SOA tower site in the MODIS products, including tile, row and column numbers.

6

7 **3. Methods**

8 3.1 Spatial integration of MODIS data

9 A possible source of discrepancy between tower-based photosynthesis and the MODIS standard GPP is the way the MODIS reflectance products are used to calculate FAPAR_{canopy} 10 and the MODIS standard GPP (Justice et al. 1998; Wolfe et al. 1998). This occurs because the 11 areal coverage of the MODIS products used in the GPP calculations are not constant over the 12 growing season and may not match the footprint of the flux tower site. The MODIS observations 13 made at multiple times over a target actually cover somewhat different ground areas due to shifts 14 in the ground track of the satellite, but are gridded into single, fixed grid cells. Footprints for off-15 nadir observations are increasingly larger and oblong in shape as VZAs increase. For example, 16 areas associated with ground targets for imagery acquired with VZAs greater than 65° are at least 17 18 nine times larger than those viewed by nadir observations (Wolfe et al. 1998). Consequently, 19 both the standard MODIS FAPAR_{canopy} and GPP products represent somewhat different, though overlapping or adjacent, areas when viewed frequently over time with different geometries. 20 21 For the current analysis, we developed a method to scale-up the MODIS land band

observations to 2500 m x 2500 m regions. We assumed that the scaled-up satellite 2500 m data
and tower flux-based data should follow a similar pattern, although they are not necessarily

1	identical. This is a reasonable assumption since the flux footprint for the site described in the
2	current study extends for at least three km in all directions from the flux tower.
3	We acquired daily MODIS data (tile H11V04) for 2001-2005. An example is given
4	(Figure 1) for MODIS daily NIR ₁ reflectances across the 5x5 grid area (where each grid cell is
5	nominally 500 m) for nadir data on day of year (DOY) 224 in 2001. The MODIS relative
6	reflectance data quality descriptor indicates that the reflectances were of high quality.
7	We used 5x5 scaled up MODIS observations to produce similar ground sectors within the
8	2.5 km x 2.5 km block area of the aspen forest (as in Fig. 1). A time series for the 5x5 block was
9	created from MODIS daily data using the following criteria: (i) only observations that fell within
10	the block were selected; (ii) an observation was excluded from consideration if the reflectance
11	quality product indicated any quality problem; (iii) observations were averaged only if their
12	geometries for view and illumination angles differed by less than five degrees; (iv) observations
13	from different swaths were not mixed; and (v) the inclusion of at least 18 of the 25 grid cells
14	were required to produce a scaled-up average observation for use in the subsequent analysis
15	(Table 1).

The NDVI time series (Tucker 1979, Eq. 5 below) over five years (2001-2005) were used to determine average core growing season length where:

18

19
$$NDVI = \frac{\rho_{NIR_1} - \rho_{red}}{\rho_{NIR_1} + \rho_{red}}.$$
 (5)

Other remote sensing indices were explored, e.g. the enhanced vegetation index (EVI; Huete *et al.* 1994) and the land surface water index (LSWI, Xiao *et al.* 2005), but NDVI gave more coherent seasonal results (data not shown). Degree 6 polynomials were used to fit average seasonal curves to the five-year data collections for the MODIS NDVI and the mid-day tower

1	GEP, respectively. Derivative analyses were applied to the fitted curves. We used the first,
2	second, and third derivatives to identify transition points in continuous data curves (Vina et al.
3	2004) to retrieve the DOY in the five-year collections when seasonal changes occurred. We
4	defined the beginning date of the growing season as the date when the second derivative reached
5	its first local extreme value during the spring green-up period. We defined the end date for the
6	growing season as that date when either the first or third derivatives attained a local extreme
7	value at the end of summer (DOY > 255 at this site), initiating autumn senescence. Based on
8	these analyses, we used NDVI for examining LUE and FAPAR dynamics for the
9	photosynthetically active period, defined as occurring between DOY 152-259. We also defined
10	the earlier and later dates for a second category representing periods of lower physiological
11	activity (DOY between 121 and 151 and DOY between 260 and 287).
12	
12 13	3.2 Estimating vegetation canopy characteristics
	3.2 Estimating vegetation canopy characteristics3.2.1 Description of the PROSAIL-2 model
13	
13 14	3.2.1 Description of the PROSAIL-2 model
13 14 15	3.2.1 Description of the PROSAIL-2 model The canopy-leaf-stem-background coupled radiative transfer model, PROSAIL-2, is an
13 14 15 16	 3.2.1 Description of the PROSAIL-2 model The canopy-leaf-stem-background coupled radiative transfer model, PROSAIL-2, is an updated version of the model used in earlier studies (Zhang <i>et al.</i> 2005, 2006). PROSAIL-2
13 14 15 16 17	 3.2.1 Description of the PROSAIL-2 model The canopy-leaf-stem-background coupled radiative transfer model, PROSAIL-2, is an updated version of the model used in earlier studies (Zhang <i>et al.</i> 2005, 2006). PROSAIL-2 resulted from the combination of the PROSPECT model used to describe leaf characteristics in
13 14 15 16 17 18	 3.2.1 Description of the PROSAIL-2 model The canopy-leaf-stem-background coupled radiative transfer model, PROSAIL-2, is an updated version of the model used in earlier studies (Zhang <i>et al.</i> 2005, 2006). PROSAIL-2 resulted from the combination of the PROSPECT model used to describe leaf characteristics in the canopy model, and the radiative transfer model, SAIL2. In the current version of SAIL2, we
13 14 15 16 17 18 19	3.2.1 Description of the PROSAIL-2 model The canopy-leaf-stem-background coupled radiative transfer model, PROSAIL-2, is an updated version of the model used in earlier studies (Zhang <i>et al.</i> 2005, 2006). PROSAIL-2 resulted from the combination of the PROSPECT model used to describe leaf characteristics in the canopy model, and the radiative transfer model, SAIL2. In the current version of SAIL2, we revised the expression for background and stem characteristics. Equations 5 & 6 in Zhang <i>et al.</i>
13 14 15 16 17 18 19 20	 3.2.1 Description of the PROSAIL-2 model The canopy-leaf-stem-background coupled radiative transfer model, PROSAIL-2, is an updated version of the model used in earlier studies (Zhang <i>et al.</i> 2005, 2006). PROSAIL-2 resulted from the combination of the PROSPECT model used to describe leaf characteristics in the canopy model, and the radiative transfer model, SAIL2. In the current version of SAIL2, we revised the expression for background and stem characteristics. Equations 5 & 6 in Zhang <i>et al.</i> (2006) were used to simulate soil and stem reflectance. In the present study, we used <i>in situ</i>

$$\rho_{back}(\lambda) = \rho_{back,\min}(\lambda) + BACK_A \bullet (\rho_{back,\max}(\lambda) - \rho_{back,\min}(\lambda))$$
(6)

$$\rho_{stem}(\lambda) = \rho_{stem,\min}(\lambda) + STEM_A \bullet (\rho_{stem,\max}(\lambda) - \rho_{stem,\min}(\lambda)), \qquad (7)$$

where λ is the spectral wavelength, ρ is reflectance, BACK_A and STEM_A are variables 3 describing reflectance values for background and stem. We used the maximum and minimum 4 reflectance values of background and stem as their upper and lower value limits. The fourteen 5 free variables used in the PROSAIL-2 for this study are summarized in Table 2. Five variables 6 were used to describe leaf characteristics: a leaf internal structure variable (N), leaf total 7 photosynthetic pigment content (C_{ab}), leaf dry matter content (C_m), leaf water thickness (C_w), and 8 leaf brown pigment content (Cbrown). The top of canopy reflectance was composed of leaf, stem 9 and background contributions. Stem fraction (SFRAC) and cover fraction (CF) were used to 10 decompose leaf, stem and background components. Refer to Zhang et al. (2005, 2006) for more 11 details on PROSAIL-2. 12

13

14 3.2.2 Description of the FAPAR_{chl} and FAPAR_{canopy} algorithm

The variables in Tab. 2 have to be estimated to calculate FAPAR_{chl} and FAPAR_{canopy} 15 using PROSAIL-2. The Metropolis algorithm, a Markov Chain Monte Carlo (MCMC) method, 16 was adopted to find solutions expressed as posterior distributions of the variables. The posterior 17 distributions of the variables are the product of their prior distributions and the likelihood 18 calculated with a Bayesian analysis (http://en.wikipedia.org/wiki/Posterior_probability). The 19 prior distributions of the input variables were assumed to be uniform (http://en.wikipedia.org/ 20 wiki/Uniform distribution (continuous)). Each MODIS reflectance observation $[\rho_{obs}]$ for the 21 seven land bands (red, NIR₁, blue, green, NIR₂, SWIR₁ and SWIR₂), and associated VZA [θ_v , in 22 degrees], relative view azimuth angle [ϕ , in degrees], and solar zenith angle [θ_s , in degrees] 23

contains some noise, although small differences in angles may be ignored. We treat each SOA
 reflectance observation as a sample of the following distribution:

3

$$\rho \sim \{\rho_{obs}(\lambda, \theta_{v}(1+3N(0,1)), \theta_{s}(1+3N(0,1)), \phi(1+3N(0,1)))\} \cdot (1+0.05N(0,1))$$
5
(8)

6 where N(0,1) is the normal distribution with a mean of zero and SD = 1.

We may use as many samples from the distribution (from equation 8) as we desire. For 7 five of the spectral bands (red, green, NIR₁, NIR₂ and SWIR₁), we calculated the log-likelihood 8 [using equations 1& 2 from Zhang et al. 2005] and then performed an acceptance test [using 9 equation 3 from Zhang et al. 2005]. A new randomly generated "proposed" value was accepted 10 only if it passed acceptance tests conducted on all five bands. The same adaptive algorithm 11 [using equation 4 from Zhang et al. 2005] was used to accelerate the speed of convergence of the 12 MCMC algorithm. With the posterior distributions of the variables (Tab. 2) now calculated, we 13 forward-simulated the fractions of APAR for canopy, leaf, and photosynthetic pigments (Zhang 14 et al. 2005, 2006) for each MODIS 5x5 scaled-up observation that met our quality rules during 15 the five-year period, using the PROSAIL-2 model. The product of LAI, CF, and leaf 16 photosynthetic pigment (μ g/cm²) describes the average photosynthetic pigment content for each 17 5x5 MODIS aspen forest observation. Similarly, the product of LAI, CF and leaf water content 18 (cm) describes the average water content of vegetation and, the product of LAI, CF and leaf dry 19 matter (g/cm²) provides the average dry matter content of vegetation in each MODIS observation. 20 21

3.2.3

Calculation of APAR _{chl} , APAR _{canopy} , L	LUE_{chb}	LUE_{canopy}	and LUE _{tower}
---	-------------	----------------	--------------------------

2	The PAR absorbed by the whole SOA canopy (APAR _{canopy}) was determined as the
3	product of PAR and the median value of the MODIS-derived FAPAR _{canopy} distribution (see
4	section above). Likewise, APAR _{chl} for only the photosynthetic pigments of the foliage
5	component was determined as the product of PAR and the median value of the MODIS-derived
6	FAPAR _{chl} distribution. For these calculations, we used the GEP and average incident canopy
7	photosynthetic photon flux density (PPFD) measured with Li-190SA PAR sensors (Licor Inc.,
8	Lincoln, NE) over the SOA tower site at 90-minute intervals centered on the satellite overpass
9	time. LUE for the foliage component of the forest at SOA (LUE _{chl} , Eq.3) was computed as the
10	ratio of GEP to APAR _{chl} . LUE for the whole SOA forest canopy (LUE _{canopy} , Eq.4) was
11	computed as the ratio of GEP to APAR _{canopy} . The LUE _{tower} (Eq. 2 above) data used in this paper
12	are cited from Krishnan <i>et al.</i> (2006) where they defined LUE_{tower} as the ratio
13	GEP _{tower} /APAR _{tower} where APAR _{tower} was estimated using Eq. (1) in Barr et al. (2007) that was
14	based on measured downwelling and upwelling PAR, overstory and understory clumping indices,
15	measured stem area indices, and the estimated daily LAI. Therefore, LUE_{canopy} and LUE_{chl} use
16	MODIS-derived FAPARs whereas LUE _{tower} derives its FAPAR equivalent from site-based
17	radiometer and structural measurements. The GEP and PAR variables, on the other hand, are the
18	same for the three LUE calculations.
19	The REGRESS function in MATLAB was used to statistically analyze regression
20	relationships between: (i) GEP and APAR _{chl} ; (ii) GEP and APAR _{canopy} ; (iii) annual average
21	LUE_{chl} ; and annual average LUE_{canopy} over the five-year core growing season period; and (iv)
22	annual average LUE_{tower} and annual average LUE_{chl} . We also used the student t-test function in
23	MATLAB to test if the slope between GEP and $APAR_{chl}$ and the slope between GEP and

APAR_{canopy} were significantly different. The same t-test function was also applied to test if the annual average LUE_{chl}, LUE_{canopy} and LUE_{tower} time series over the entire five-year period were significantly different.

- 4
- 5

3.2.4 Controls on inter-annual LUE_{chl} and LUE_{canopy}

To determine the controls on the inter-annual variation of LUE, we analyzed the 6 7 relationship between LUE_{chl}, LUE_{canopy}, soil water content (SWC), and precipitation as well as the three factors that determine canopy chlorophyll concentration, i.e., average leaf chlorophyll 8 concentration, leaf area index (LAI), and cover fraction (CF). Soil water content was measured 9 with eight time domain reflectometry (TDR) probes (Moisture Point type B, Gabel Corp., 10 Victoria, Canada) placed at 10 m intervals with measurements at depths of 0-15, 15-30, 30-60, 11 60-90, and 90-120 cm, although SWC values were used only for the 0-30 cm zone because this is 12 where 90% of the roots are found (Barr et al. 2007). Rain precipitation was measured with a 13 Geonor T200 accumulation rain gauge (Geonor Inc, Milford, PA) supplemented by a CS700 14 tipping bucket rain gauge (Campbell Scientific Inc, Edmonton, AB). The half-hourly 15 precipitation measurements at the SOA tower site in 2001-2005 were summed into monthly rain 16 precipitation totals, from which annual values were 235 mm, 285.8 mm, 261 mm, 667 mm and 17 614 mm for 2001 - 2005, respectively. 18

19

20 4. Results

21 4.1 PROSAIL-2 Derived Canopy Variables including APAR_{chl} and APAR_{canopy}

The posterior distributions of SOA canopy variables from the PROSAIL-2 model display seasonal variation, as shown for nine of the seventeen MODIS daily 5x5 observations in 2005

1	(Table 3). The other observations had similar distributions as those shown in Tab. 3 so they are
2	not presented. Several canopy variables are shown: LAI, CF, total photosynthetic pigment
3	content, water content, dry matter content, FAPAR _{canopy} , and FAPAR _{chl} . Differences in total
4	chlorophyll concentration were related to changes in all three of the factors used in its calculation,
5	i.e., average leaf chlorophyll concentration, LAI, and CF. The considerably higher FAPAR _{canopy}
6	(0.47 - 0.87), as compared to FAPAR _{chl} $(0.03 - 0.70)$, results from PAR absorption primarily by
7	non-photosynthetic canopy components. The average $APAR_{chl}$ at SOA over the five-year period
8	was roughly 65% of $APAR_{canopy}$ (slope of the all-data relationship in Fig. 2). We also found that
9	the average ratios for APAR _{chl} : APAR _{canopy} were different between the core growing season,
10	DOY 152 – 259 (\blacklozenge , slope = 0.71); and the combined early and late periods of the season, DOY
11	<152 and DOY>259 (Δ , slope = 0.34) (Fig. 2). The correlation between APAR _{chl} and
12	APAR _{canopy} showed the highest correlation ($r = 0.87$) during the core growing season, DOY 152-
13	259 (Fig. 2).

15 4.2 LUE_{chl} and LUE_{canopy} Over the Five-Year Period

Average values for LUE_{canopy} and LUE_{chl} over the five-year period (2001-2005) were 0.0155 and 0.0241 µmol C µmol⁻¹ PPFD, respectively, as determined from the slopes of the GEP:APAR relationships (Figure 3 for APAR_{canopy}; Figure 4 for APAR_{chl}, p<0.0001). For the entire study period (DOY ranging from 121 to 287), there was a stronger correlation between GEP and APARchl, (Fig. 4, $r^2 = 0.78$) compared GEP:APAR_{canopy} (Fig. 3, $r^2 = 0.33$). The 95% confidence intervals for the five-year average LUE_{canopy} and LUE_{chl} did not overlap, i.e., they ranged from 0.0141 to 0.0169 and from 0.0229 to 0.0253, respectively.

1	For the core growing season only (DOY 152 – 259), the five-year average values of
2	LUE_{canopy} and LUE_{chl} were 0.0173 and 0.0243 µmol C µmol ⁻¹ PPFD, respectively (Figure 3 for
3	APAR _{canopy} ; Figure 4 for APAR _{chl} , p<0.0006). Once again, the 95% confidence intervals for the
4	five-year average LUE_{canopy} and LUE_{chl} did not overlap, i.e., they ranged from 0.0163 to 0.0183
5	and from 0.0230 to 0.0256, respectively.
6	$APAR_{chl}$ and LUE_{chl} captured more seasonal variation than their whole canopy
7	counterparts (APAR _{canopy} , LUE _{canopy}) (Figs. 3 and 4). For example, during the early and late
8	growing season, average LUE_{chl} over five years was 0.0208 µmol C µmol ⁻¹ PPFD which was
9	lower than the core growing season value of 0.0243 μ mol C μ mol ⁻¹ PPFD (Fig. 4).
10	
11	4.3 Inter-Annual Variability in LUE
12	There were also inter-annual variations for $APAR_{chl}$ and $APAR_{canopy}$ (mean ± SE, the
13	standard error) with consistently and significantly lower values for APAR _{chl} compared to
14	APAR _{canopy} during the core growing season (Figure 5) (p<0.005). Consequently, LUE_{chl} was
15	consistently and significantly higher than LUE_{canopy} (Figure 6) (p<0.005) and LUE_{chl} in 2004-
16	2005 was higher than for the three earlier years during the growing season. The effect of the
17	2003 drought was apparent, such that the average LUE_{chl} and LUE_{canopy} in 2004 were much
18	higher than in 2003. The annual means of core growing season LUE_{chl} (µmol C µmol ⁻¹ PPFD)
19	$(\pm SE)$ were: 0.0242 \pm 0.0012 (2001), 0.0245 \pm 0.0015 (2002); 0.0225 \pm 0.0018 (2003); 0.0310 \pm
20	0.0022 (2004); and 0.0267 \pm 0.0019 (2005) (Tab. 4). The maximum LUE (ϵ_{max}) value for
21	broadleaf deciduous forests set by the biome look-up table of the MODIS photosynthesis model
22	is 0.0203 μ mol C μ mol ⁻¹ PPFD (Heinsch et al., 2003). Four of the five annual LUE _{canopy} values

that we calculated were lower than ε_{max} . However, the 2004 LUE_{canopy} was much higher than the MODIS ε_{max} value (Fig. 6).

3	Average LAI, CF, canopy chlorophyll concentration, FAPAR _{canopy} , FAPAR _{chl} , LUE _{canopy}
4	and LUE _{chl} during core growing season ($152 \le DOY \le 259$) for each individual year are shown in
5	Table 4. Inter-annual variations of canopy chlorophyll concentration were influenced by all the
6	three factors ($r^2 = 0.91$ for leaf chlorophyll concentration; $r^2 = 0.32$ for LAI; and $r^2 = 0.49$ for CF).
7	Average annual canopy chlorophyll concentrations were correlated with both $FAPAR_{chl}$ ($r^2 =$
8	0.81) and $APAR_{chl}$ (r ² = 0.63).
9	
10	4.4. Comparisons with LUE _{tower}
11	Whereas LUE_{canopy} differed significantly from LUE_{tower} for each of the five years
12	(p \leq 0.001), LUE _{chl} was essentially the same as LUE _{tower} (p \geq 0.47, Figure 6).
13	
13 14	5. Discussion
	 Discussion LUE_{chl} captured more seasonal (Tab. 3) and inter-annual variation than LUE_{canopy} and
14	
14 15	LUE_{chl} captured more seasonal (Tab. 3) and inter-annual variation than LUE_{canopy} and
14 15 16	LUE_{chl} captured more seasonal (Tab. 3) and inter-annual variation than LUE_{canopy} and provided an improved overall relationship to GEP (Figs. 3 and 4). Krishnan <i>et al.</i> (2006) reported
14 15 16 17	LUE_{chl} captured more seasonal (Tab. 3) and inter-annual variation than LUE_{canopy} and provided an improved overall relationship to GEP (Figs. 3 and 4). Krishnan <i>et al.</i> (2006) reported that the annual average LUE_{tower} at the SOA tower for 2001-2005 was 0.0229 – 0.0302 µmol C
14 15 16 17 18	LUE _{chl} captured more seasonal (Tab. 3) and inter-annual variation than LUE _{canopy} and provided an improved overall relationship to GEP (Figs. 3 and 4). Krishnan <i>et al.</i> (2006) reported that the annual average LUE _{tower} at the SOA tower for 2001-2005 was $0.0229 - 0.0302 \mu$ mol C μ mol ⁻¹ PPFD. Their <i>in situ</i> average LUE _{tower} estimate matched well with our average MODIS-
14 15 16 17 18 19	LUE _{chl} captured more seasonal (Tab. 3) and inter-annual variation than LUE _{canopy} and provided an improved overall relationship to GEP (Figs. 3 and 4). Krishnan <i>et al.</i> (2006) reported that the annual average LUE _{tower} at the SOA tower for 2001-2005 was $0.0229 - 0.0302 \mu$ mol C μ mol ⁻¹ PPFD. Their <i>in situ</i> average LUE _{tower} estimate matched well with our average MODIS- derived LUE _{chl} (0.0229 – 0.0302 vs. 0.0225 – 0.0310 µmol C µmol ⁻¹ PPFD from Fig. 4) but not
14 15 16 17 18 19 20	LUE _{chl} captured more seasonal (Tab. 3) and inter-annual variation than LUE _{canopy} and provided an improved overall relationship to GEP (Figs. 3 and 4). Krishnan <i>et al.</i> (2006) reported that the annual average LUE _{tower} at the SOA tower for 2001-2005 was $0.0229 - 0.0302 \mu$ mol C μ mol ⁻¹ PPFD. Their <i>in situ</i> average LUE _{tower} estimate matched well with our average MODIS- derived LUE _{chl} (0.0229 – 0.0302 vs. 0.0225 – 0.0310 µmol C µmol ⁻¹ PPFD from Fig. 4) but not with the MODIS-derived LUE _{canopy} over five growing seasons. Furthermore, the <i>in situ</i> tower-

The annual LUE_{canopy} values substantially underestimated the tower-based estimates 1 (Krishnan et al. 2006) (Fig. 6). It is interesting to note that the five-year average LUE_{canopy} for 2 the mid-growing season period was higher than LUE_{canopy} for the early and late season periods 3 (Fig. 3). In comparison, the five-year average LUE_{chl} for all data was consistent with LUE_{chl} for 4 the mid-growing season period, and also close to LUE_{chl} for the early and late season periods 5 (Fig. 4). Additionally, the maximum LUE value (ε_{max}) from the MODIS biome look-up table for 6 7 broadleaf deciduous forests (Heinsch et al. 2003) tended to be significantly lower than both LUE_{chl} and LUE_{tower} (Fig. 6). This study has demonstrated that (1) LUE_{chl} values are more 8 comparable to ground-based observations of LUE_{tower} than LUE_{canopy}; (2) APAR_{chl} values are 9 more comparable to ground-based observations of APAR_{tower} when the sky is clear than 10 APAR_{canopy}; (3) FAPAR_{chl} values derived from MODIS observations are more realistic and 11 useful for estimating of APAR_{tower} when the sky is clear than FAPAR_{canopy} values. The 12 conclusions are supported by measurements and simulations of photosynthetic vs. non-13 14 photosynthetic vegetation by Chen et al. (2006) at other flux sites in the region. Consequently, we accept the hypothesis that $LUE_{chl} = LUE_{tower}$, whereas we reject the other two hypotheses that 15 $LUE_{canopy} = LUE_{chl}$ or that $LUE_{canopy} = LUE_{tower}$. 16

A prolonged three-year drought began in late summer of 2001 and the most severe conditions occurred in 2003 (Barr *et al.* 2007), which also produced the lowest LUE_{chl} (Fig. 6). There were eleven consecutive months prior to August 2003 when precipitation was <50 mm per month. The water table depth decreased from 3 m in 2001 to 4 m in 2003 (Barr *et al.* 2007). Low precipitation in 2001 through 2003 caused soil water content in the shallow layer (0-0.15 m) to begin to drop in August 2001 and it kept dropping through 2002 and 2003. The lowest soil water content occurred in 2003 and was one-third lower than the pre-drought mean value and was close to the permanent wilting point. Surface conductance declined during the drought years and reached its lowest value in 2003 (Krishnan *et al.* 2006). Figure 7 compares the average LUE_{chl} and LUE_{canopy} during DOY 152-259 of each year (2001 - 2005) with the cumulative rain precipitation during that time period. This analysis suggests that the cumulative precipitation during 2001-2005 may have had a significant influence on LUE. Increased rain precipitation in 2004 and 2005 recharged the soil and increased LUE and our MODIS-based LUE_{chl} estimates were sensitive to this phenomenon.

FAPAR_{chl} and LUE_{chl} are more physiologically realistic ways of quantifying the PAR 8 9 absorbed and used for photosynthesis. These are integrative measures that inherently account for some of the impacts of mid- to long-term stresses since they also reflect changes in leaf area and 10 11 cover fraction. For example, inter-annual variability of total canopy chlorophyll concentration was influenced by changes in average leaf chlorophyll concentration, LAI and CF (Table 4). 12 The 2003 drought lowered all three of these factors at our study area relative to 2001 and 2002. 13 Thus, chlorophyll-based measures of APAR and LUE have the potential to more directly account 14 15 for environmental limitations and thus reduce the impact of the uncertainty in estimating temperature, vapor pressure deficit, and soil moisture for specific pixels. Waring et al. (1995) 16 found a strong correlation between upper canopy leaf chlorophyll concentration of the major 17 hardwood species and maximum light use efficiency at the primarily deciduous Harvard Forest 18 flux site. Their finding supports the idea of a direct link between pigment concentration and 19 LUEtower for deciduous forests. Ollinger et al. (2008) and Kergoat et al. (2008) showed a 20 significant positive relationship between whole canopy nitrogen concentration and canopy 21 maximum LUE but the relationship between canopy nitrogen and chlorophyll was not described 22 and they did not examine inter-annual or seasonal variability at any of their sites. Their results 23

lead us to believe that an inter-site analysis based on APAR_{chl} and LUE_{chl} using our methodology
 would yield even stronger relationships.

Chlorophyll-based measures of APAR and LUE, however, will not account for 3 limitations due to short-term environmental extremes so modulation of light use efficiency by 4 environmental stresses will still need to be considered. Furthermore, photosynthesis of evergreen 5 conifer forests are less sensitive to changes in chlorophyll than are broadleaf forests as evidenced 6 by the continuous green color of conifer forests even during the coldest periods of winter. As 7 well, forests that have attained maximum height are subject to significant hydraulic and stomatal 8 limitations (Ryan et al. 2004, 1997) that may or may not be reflected in their total canopy 9 chlorophyll concentration. Brodribb and Feild (2000) demonstrated a highly significant 10 correlation between hydraulic conductivity, maximum photosynthetic capacity, and quantum 11 yield 23 rain forest species. 12

To our knowledge, this study represents the first time that LUE_{chl} has been estimated by 13 linking tower flux data, a biophysical radiative transfer model, and satellite spectral observations. 14 Our predicted values for LUE_{chl} agree well with *in situ* data in respect to both amplitude and 15 seasonal phase; and our modeled LUE_{chl} successfully described the actual dynamics captured by 16 tower fluxes. We believe that it could be useful to couple this type of FAPAR_{chl} approach into 17 18 regional/global carbon cycle models, land surface process models and general circulation models. 19 This paper has demonstrated some of the possible benefits of using FAPAR_{chl} as an operational data product for carbon cycle modeling. The hyperspectral satellite sensors that are 20 21 in orbit (e.g., EO-1/Hyperion) or currently under development (e.g., HyspIRI) could help us obtain even more robust estimates of FAPAR_{chl} due to the greater sensitivity of these sensors to 22 pigment concentrations and other biochemical properties of foliage (Coops et al. 2002). A 23

1	narrower field of view could also be helpful for resolving chlorophyll dynamics at scales more
2	representative (e.g., <100 m) of the spatial structure typical of most forest stands.
3	
4	6. Acknowledgments
5	This project was supported by the NASA Carbon Cycle Science Program (Dr. Diane Wickland,
6	Program Manager) under the auspices of the North American Carbon Program. Support was
7	provided to the FCRN and BERMS programs through several Canadian funding sources
8	(Environment Canada, the Canadian Forest Service, NSERC, CFCAS, BIOCAP, Action Plan
9	2000 and PERD).
10	
11	7. References
12	Barr, A.G., Black, T.A., Hogg, E.H., Griffis, T.J., Morgenstern, K., Kljun, N., Theede, A., &
13	Nesic, Z. (2007). Climatic controls on the carbon and water balances of a boreal aspen
14	forest, 1994-2003. Global Change Biology, 13, 561-576.
15	Brodribb, T.J. & Feild, T.S. (2000). Stem hydraulic supply is linked to leaf photosynthetic
16	capacity: evidence from New Caledonian and Tasmanian rainforests. Plant, Cell and
17	Environment 23, 1381-1388
18	Coops, N.C., Smith, ML., Martin, M.E., Ollinger, S.V., & Held, A. (2002). Predicting eucalypt
19	biochemistry from HYPERION and HYMAP imagery. IGARSS'02, 2, 790 - 792
20	Gower, S.T., Kucharik, C.J., & Norman, J.M. (1999). Direct and indirect estimation of leaf area
21	index, fAPAR and net primary production of terrestrial ecosystems. Remote Sensing of
22	Environment, 70, 29-51

1	Griffis, T.J., Black, T.A., Morgenstern, K., Barr, A.G., Nesic, Z., Drewitt, G.B., Gaumont-Guay,
2	D., & McCaughey, J.H. (2003). Ecophysiological controls on the carbon balances of
3	three southern boreal forests. Agricultural and Forest Meteorology, 117, 53-71
4	Heinsch, F.A., M. Reeves, P. Votava, S. Kang, C. Milesi, M. Zhao, J. Glassy, W.M. Jolly, R.
5	Loehman, C.F. Bowker, J. Kimball, R. Nemani, S.W. Running. (2003). User's Guide:
6	GPP and NPP (MOD17A2/A3) Products. NASA MODIS Land Algorithm. Version 2.0.
7	NASA Algorithm Theoretical Basis Document (ATBD).
8	Huete, A., Justice, C., & Liu, H. (1994). Development of Vegetation and Soil Indexes for Modis-
9	Eos. Remote Sensing of Environment, 49, 224-234
10	Justice, C.O., Vermote, E., Townshend, J.R.G., Defries, R., Roy, D.P., Hall, D.K., Salomonson,
11	V.V., Privette, J.L., Riggs, G., Strahler, A., Lucht, W., Myneni, R.B., Knyazikhin, Y.,
12	Running, S.W., Nemani, R.R., Wan, Z.M., Huete, A.R., van Leeuwen, W., Wolfe, R.E.,
13	Giglio, L., Muller, J.P., Lewis, P., & Barnsley, M.J. (1998). The Moderate Resolution
14	Imaging Spectroradiometer (MODIS): Land remote sensing for global change research.
15	IEEE Transactions on Geoscience and Remote Sensing, 36, 1228-1249
16	Kergoat, L., Lafont, S., Arneth, A., Le Dantec, V., & Saugier, B. (2008). Nitrogen controls
17	plant canopy light-use efficiency in temperate and boreal ecosystems. Journal of
18	Geophysical Research, 113, G04017, doi:10.1029/2007JG000676.
19	Krishnan, P., Black, T.A., Grant, N.J., Barr, A.G., Hogg, E.H., Jassal, R.S., & Morgenstern, K.
20	(2006). Impact of changing soil moisture distribution on net ecosystem productivity of a
21	boreal aspen forest during and following drought. Agricultural and Forest Meteorology,
22	139, 208-233

1	Landsberg, J.J., & Waring, R.H. (1997). A generalized model of forest productivity using
2	simplified concepts of radiation-use efficiency, carbon balance and partitioning. Forest
3	Ecology and Management, 95, 209–228
4	Law, B.E., Waring, R.H., Anthoni, P.M., & Aber, J.D. (2000). Measurements of gross and net
5	ecosystem productivity and water vapour exchange of a Pinus ponderosa ecosystem, and
6	an evaluation of two generalized models. Global Change Biology, 6, 155-168
7	Medlyn, B.E. (1998). Physiological basis of the light use efficiency model. Tree Physiology, 18,
8	167-176
9	Mohr, H., Schopfer, P., Lawlor, G., & Lawlor, D.W. (Eds.) (1995). Plant Physiology: Springer
10	Monteith, J.L. (1972). Solar-Radiation and productivity in tropical ecosystems. Journal of
11	Applied Ecology, 9, 747-766
12	Monteith, J.L. (1977). Climate and efficiency of crop production in Britain. Philosophical
13	Transaction of the Royal Society of London B: Biological Sciences, 281, 271-294
14	Ollinger, S.V., Richardson, A.D., Martin, M.E., Hollinger, D.Y., Frolking, S.E., Reich, P.B.,
15	Plourde, L.C., Katul, G.G., Munger, J.W., Oren, R.O., Smith, ML., Paw U, K.T.,
16	Bolstad, P.V., Cook, B.D., Day, M.C., Martin, T.A., Monson, R.K., & Schmid, H.P.
17	(2008). Canopy nitrogen, carbon assimilation, and albedo in temperate and boreal forests:
18	Functional relations and potential climate feedbacks. Proceedings, National Academy of
19	Sciences, 105, 19336-19341.
20	Prince, S.D., Goetz, S.J., & Goward, S.N. (1995). Monitoring Primary Production from Earth
21	Observing Satellites. Water Air and Soil Pollution, 82, 509-522
22	Prince, S.D., & Goward, S.N. (1995). Global primary production: A remote sensing approach.
23	Journal of Biogeography, 22, 815-835

1	Prince, S.D., & Goward, S.N. (1996). Evaluation of the NOAA/NASA Pathfinder AVHKR Land
2	Data Set for global primary production modelling. International Journal of Remote
3	Sensing, 17, 217-221
4	Prince, S.D., Goward, S.N., Goetz, S., & Czajkowski, K. (2000). Interannual Atmosphere-
5	Biosphere Variation: Implications for observation and modeling. Journal of Geophysical
6	Research-Atmospheres, 105, 20055-20063
7	Ruimy, A., Dedieu, G., & Saugier, B. (1996a). TURC: A diagnostic model of continental gross
8	primary productivity and net primary productivity. Global Biogeochemical Cycles, 10,
9	269-285
10	Ruimy, A., Kergoat, L., Field, C.B., & Saugier, B. (1996b). The use of CO2 flux measurements
11	in models of the global terrestrial carbon budget. Global Change Biology, 2, 287-296
12	Ruimy, A., Saugier, B., & Dedieu, G. (1994). Methodology for estimation of terrestrial net
13	primary production from remotely sensed data. Journal of Geophysical Research -
14	Atmospheres, 99, 5263-5283
15	Running, S., Nemani, R., Heinsch, F., Zhao, M., Reeves, M., & Hashimoto, H. (2004). A
16	continuous satellite-derived measure of global terrestrial primary production. BioScience,
17	54, 547-560
18	Running, S.W., Baldocchi, D.D., Turner, D.P., Gower, S.T., Bakwin, P.S., & Hibbard, K.A.
19	(1999a). A global terrestrial monitoring network integrating tower fluxes, flask sampling,
20	ecosystem modeling and EOS satellite data. Remote Sensing of Environment, 70, 108-127
21	Running, S.W., Justice, C.O., Salomonson, V., Hall, D., Barker, J., Kaufmann, Y.J., Strahler,
22	A.H., Huete, A.R., Muller, J.P., Vanderbilt, V., Wan, Z.M., Teillet, P., & Carneggie, D.

1	(1994). Terrestrial Remote-Sensing Science and Algorithms Planned for Eos Modis.
2	International Journal of Remote Sensing, 15, 3587-3620
3	Running, S.W., Nemani, R., Glassy, J.M., & Thornton, P.E. (1999b). MODIS daily
4	photosynthesis (PSN) and annual net primary production (NPP) product (MOD17),
5	Algorithm Theoretical Basis Document, Version 3.0
6	Running, S.W., Thornton, P.E., Nemani, R., & Glassy, J.M. (2000). Global terrestrial gross and
7	net primary productivity from the Earth Observing System. In O.E. Sala, R.B. Jackson,
8	H.A. Mooney & R.W. Howarth (Eds.), Methods in Ecosystem Science (pp. 44-57). New
9	York: Springer Verlag
10	Ryan M.G., Binkley D., Fownes J.H. (1997) Age-related decline in forest productivity: pattern
11	and process. Advances in Ecological Research, 27, 213-262.
12	Ryan M.G., Binkley D., Fownes J.H., Giardina C.P., Senock, R.S. (2004). An experimental test
13	of the causes of forest growth decline with stand age. Ecological Monographs, 74, 393-
14	414.
15	Sellers, P.J., Los, S.O., Tucker, C.J., Justice, C.O., Dazlich, D.A., Collatz, G.J., & Randall, D.A.
16	(1996a). A revised land surface parameterization (SiB2) for atmospheric GCMs .2. The
17	generation of global fields of terrestrial biophysical parameters from satellite data.
18	Journal of Climate, 9, 706-737
19	Sellers, P.J., Randall, D.A., Collatz, G.J., Berry, J.A., Field, C.B., Dazlich, D.A., Zhang, C.,
20	Collelo, G.D., & Bounoua, L. (1996b). A revised land surface parameterization (SiB2)
21	for atmospheric GCMs .1. Model formulation. Journal of Climate, 9, 676-705
22	Tucker, C.J. (1979). Red and Photographic Infrared Linear Combinations for Monitoring
23	Vegetation. Remote Sensing of Environment, 8, 127-150

1	Turner, D.P., Ollinger, S., Smith, M.L., Krankina, O., & Gregory, M. (2004). Scaling net
2	primary production to a MODIS footprint in support of Earth observing system product
3	validation. International Journal of Remote Sensing, 25, 1961-1979
4	Turner, D.P., Ritts, W.D., Cohen, W.B., Gower, S.T., Zhao, M.S., Running, S.W., Wofsy, S.C.,
5	Urbanski, S., Dunn, A.L., & Munger, J.W. (2003). Scaling Gross Primary Production
6	(GPP) over boreal and deciduous forest landscapes in support of MODIS GPP product
7	validation. Remote Sensing of Environment, 88, 256-270
8	Turner, D.P., Ritts, W.D., Zhao, M., Kurc, S.A., Dunn, A.L., Wofsy, S., Small, E., & Running,
9	S.W. (2006). Assessing interannual variation in MODIS-based estimates of gross primary
10	production. Ieee Computational Science & Engineering, 44, 1899-1907
11	Vina, A., A. Gitelson, D. Rundquist, G. Keydan, B. Leavitt, and J. Schepers (2004). Monitoring
12	Maize (Zea mays L.) Phenology with Remote Sensing. Agronomy Journal 96:1139-1147.
13	Waring, R.H., Law, B.E., Goulden, M.L., Bassow, S.L., McCreight, R.W., Wofsy, S.C., Bazzaz,
14	F.A. (1995). Scaling gross ecosystem production at Harvard Forest with remote sensing:
15	a comparison of estimates from a constrained quantum-use efficiency model and eddy
16	correlation. Plant, Cell, Environment 18, 1201-1213.
17	Wolfe, R., Roy, D., & Vermote, E. (1998). The MODIS land data storage, gridding and
18	compositing methodology: Level 2 Grid. IEEE Transactions on Geosciences and Remote
19	Sensing, 36, 1324-1338
20	Xiao, X.M., Zhang, Q.Y., Hollinger, D., Aber, J., & Moore, B. (2005). Modeling gross primary
21	production of an evergreen needleleaf forest using modis and climate data. Ecological
22	Applications, 15, 954-969

1	Yoder, B.J., & Waring, R.H. (1994). The Normalized Difference Vegetation Index of Small
2	Douglas-Fir Canopies with Varying Chlorophyll Concentrations. Remote Sensing of
3	Environment, 49, 81-91
4	Zhang, Q., Xiao, X., Braswell, B., Linder, E., Ollinger, S., Smith, M.L., Jenkins, J.P., Baret, F.,
5	Richardson, A.D., Moore, B., & Minocha, R. (2006). Characterization of seasonal
6	variation of forest canopy in a temperate deciduous broadleaf forest, using daily MODIS
7	data Remote Sens. Environ., 105, 189-203
8	Zhang, Q.Y., Xiao, X.M., Braswell, B., Linder, E., Baret, F., & Moore, B. (2005). Estimating
9	light absorption by chlorophyll, leaf and canopy in a deciduous broadleaf forest using
10	MODIS data and a radiative transfer model. Remote Sens. Environ., 99, 357-371
11	
12	
13	
14	

1 2	List of abbi	reviations and acronyms
3	BERMS -	Boreal Ecosystem Research and Monitoring Sites
4	BOREAS -	BOReal Ecosystem-Atmosphere Study
5	BRDF -	bidirectional reflectance distribution function
6	chl -	chlorophyll
7	EVI -	enhanced vegetation index
8	FAPAR -	fraction of absorbed photosynthetically active radiation
9	FCRN -	Fluxnet-Canada Research Network
10	LSWI -	land surface water index
11	LUE -	light use efficiency
12	MCMC -	Markov Chain Monte Carlo
13	MODIS -	MODerate resolution Imaging Spectroradiometer
14	NDVI -	normalized difference vegetation index
15	SOA -	the Southern Old Aspen tower site
16		
17		
18		
19		
20		
21		
22		
23		
24		

Table 1. Days of useable daily 5x5 MODIS observations over the Southern Old Aspen site (SOA)

Year	Day of Year (DOY)
2001	144, 153, 155, 185, 189, 190, 215, 216, 218, 222,
n=22	224, 228, 233, 236, 245, 247, 249, 258, 261, 265,
	272, 278
2002	156, 174, 177, 179, 188, 190, 193, 197, 216, 232,
n=16	235, 236, 239, 241, 261, 268
2003	139, 145, 148, 155, 166, 168, 197, 214, 226, 230,
n=12	237, 246
2004	182, 199, 206, 207, 208, 212, 217, 226, 247
n=9	
2005	121, 122, 123, 128, 132, 150, 160, 173, 196, 219,
n=17	233, 244, 245, 247, 249, 285, 287

during 2001-2005 (n=76)

	Variable	Description	Unit	Search range
Biophysical /biochemical	PAI	Plant area index, i.e., leaf +stem area index	m^2/m^2	1 – 7.5
variables	SFRAC	Stem fraction		0-1
	CF	Cover fraction: area of land covered by vegetation/total area of land		0.5 –1
	C _{ab}	Leaf photosynthetic pigments including chlorophyll a+b and carotenoids	µg/cm ²	0 - 150
	N	Leaf structure variable: measure of the internal structure of the leaf		1.0 - 4.5
	Cw	Leaf equivalent water thickness	cm	0.001 - 0.15
	Cm	Leaf dry matter content	g/cm ²	0.001 - 0.04
	C _{brown}	Leaf brown pigment content		0.00001 - 8
	LFINC	Mean leaf inclination angle	degree	10 - 89
	STINC	Mean stem inclination angle	degree	10 - 89
	LFHOT	Leaf BRDF variable: length of leaf/ height of vegetation canopy		0 – 0.9
	STHOT	Stem BRDF variable: length of stem/ height of vegetation canopy		0 - 0.9
	STEMA	Stem reflectance variable		0.0 - 1.0
	BACK _A	Background reflectance variable		0.0 - 1.0

Table 2. A list of variables in the PROSAIL-2 model and the search ranges used for inversion.

Table 3. Median values for photosynthetic pigments, water content, and dry matter at the gridcell level. LAI, CF, FAPAR_{canopy} and FAPAR_{chl} were estimated using PROSAIL-2 from daily 5x5MODIS observations over the Southern Old Aspen site (SOA) in 2005. Note that the grid cellestimate = leaf level estimate*LAI*CF (where "estimate" refers to either photosyntheticpigments, water content, or dry matter, respectively)

DOY	LAI	CF	Photosynthetic	Water	Dry	FAPAR _{canopy}	FAPAR _{chl}
in 2005			pigment	content	matter	0-1	0-1
			$(\mu g/cm^2)$	(cm)	(g/cm^2)		
132	0.27	0.60	3.75	0.0005	0.0004	0.470	0.030
150	1.45	1.00	45.19	0.0050	0.0107	0.627	0.363
160	3.30	1.00	103.07	0.0311	0.0454	0.796	0.529
173	2.87	0.96	159.73	0.0326	0.0365	0.775	0.674
196	3.41	0.95	188.30	0.0345	0.0381	0.805	0.703
219	3.89	0.94	170.32	0.0385	0.0459	0.816	0.659
233	3.83	0.97	148.91	0.0417	0.0472	0.872	0.601
247	2.73	0.99	121.08	0.0450	0.0548	0.853	0.497
285	0.10	0.86	2.55	0.0005	0.0004	0.650	0.034

Table 4. Annual averages ($152 \le DOY \le 259$) of the median grid cell values of LAI, CF,

photosynthetic pigment concentration, $FAPAR_{canopy}$, $FAPAR_{chl}$, LUE_{canopy} and LUE_{chl} for the area

Year	LAI	CF	Photosynthetic	FAPAR _{canopy}	FAPAR _{chl}	LUE _{canopy}	LUE _{chl}
			pigment	0-1	0-1	(µ mol C	(µ mol C
		_	$(\mu g/cm^2)$			μ mol ⁻¹ PPFD)	μ mol ⁻¹ PPFD)
2001	3.49	0.98	175.73	0.856	0.652	0.0183	0.0242
2002	3.35	0.96	161.32	0.815	0.617	0.0183	0.0245
2003	3.23	0.94	123.68	0.794	0.582	0.0163	0.0225
2004	3.54	0.96	140.32	0.811	0.623	0.0244	0.0310
2005	3.14	0.97	135.56	0.820	0.589	0.0192	0.0267

around the Old Aspen flux site.

Figure Captions

- Fig. 1. Nadir MODIS NIR₁ reflectances (VZA < 5⁰) for the 5 x 5 area (2.5 km x 2.5 km) on DOY 224 in 2001. The central grid cell covers the Southern Old Aspen [SOA] tower site in Canada.
- Fig. 2. The relationship of APAR_{chl} = [(90 min PPFD)*FAPAR_{chl}] to APAR_{canopy} = [(90 min PPFD)*FAPAR_{canopy}] for 2001-2005. Solid diamonds (\blacklozenge) indicate values retrieved during the core growing season (DOY:152-259; r = 0.87; APAR_{chl} = 0.713 *APAR_{canopy}). Early (DOY<152) and late (DOY>259) season values were combined and are indicated with open triangles (Δ) (r = 0.55; APAR_{chl} = 0.339* APAR_{canopy}). For all values (dashed line), r = 0.72 and APAR_{chl} = 0.649*APAR_{canopy}.
- Fig. 3. The relationship of GEP to $APAR_{canopy} = [(90 \text{ min PPFD})*FAPAR_{canopy}]$ in 2001-2005. Solid diamonds (\blacklozenge) are data from during the core growing season (DOY 152-259). Open triangles (Δ) indicate values obtained before (DOY<152) or after (DOY>259). The apparent LUE_{canopy}, the slope of the relationship (0.0155 µmol C µmol⁻¹ PPFD, r² = 0.33), is similar to the core growing season value (0.0173) but is considerably higher than for days having low GEP in the spring and fall.

Fig. 4. The relationship of GEP to $APAR_{chl} = [(90 \text{ min PPFD})*FAPAR_{chl}]$ in 2001-2005. Solid diamonds (\blacklozenge) are data from during the core growing season (DOY 152-259, GEP = 0.0243*APAR_{chl}, r² = 0.63). Open triangles (Δ) indicate values obtained before (DOY<152) or

after (DOY>259) (GEP = $0.0208*APAR_{chl}$, $r^2 = 0.64$). The apparent LUE_{chl}, i.e., the slope of the relationship for all values, is $0.0241 \,\mu\text{mol} \,\text{C} \,\mu\text{mol}^{-1}$ PPFD ($r^2 = 0.78$).

- Fig. 5. Comparison of the annual means ± SE for APAR_{chl} (♦) and APAR_{canopy} (□) during the five-year period (2001-2005) for the growing season between DOY=152-259. APAR_{chl} was significantly lower than APAR_{canopy} in every year, averaging 235 µmol PPFD m⁻²s⁻¹ less than APAR_{canopy}. APAR_{tower} is not included in this figure because unlike APAR_{chl} and APAR_{canopy}, it was obtained under all sky conditions, i.e., both clear and cloudy.
- Fig. 6. Comparison of the annual means \pm SE for MODIS-derived LUE_{chl}(\blacklozenge) and LUE_{canopy} (\Box) during the five-year period (2001-2005) for the core growing season between DOY =152-259. LUE_{chl} was significantly higher than LUE_{canopy} in every year, averaging 0.007 µmol C µmol⁻¹ PPFD higher. Annual LUE_{tower} values (Krishnan *et al.*, 2006) (Δ) agree well with our LUE_{chl}, falling within the SE range in 4 of 5 years. The maximum LUE (ε_{max}) for broadleaf forests used by the MODIS PSN model is shown as a horizontal dashed line.
- **Fig. 7**. Linear relationship between annual average LUE as a function of the cumulative precipitation for the core growing season (DOY=152–259) in 2001-2005.

0.0110	0.0014	0.071	0.0004	0.0500
0.2112	0.2614	0.271	0.3094	0.2589
0.2916	0.3198	0.3079	0.3292	0.3335
0.3198	0.3294	0.338	0.3382	0.2982
0.3342	0.3318	0.2683	0.2982	0.2738
0.0042	0.0010	0.2000	0.2302	0.2700
0.3211	0.2683	0.301	0.2784	0.1331

Fig. 1. Nadir MODIS NIR, reflectances (VZA < 5°) for the 5 x 5 area (2.5 km x 2.5 km) on DOY 224 in 2001. The central grid cell covers the Southern Old Aspen [SOA] tower site in Canada.

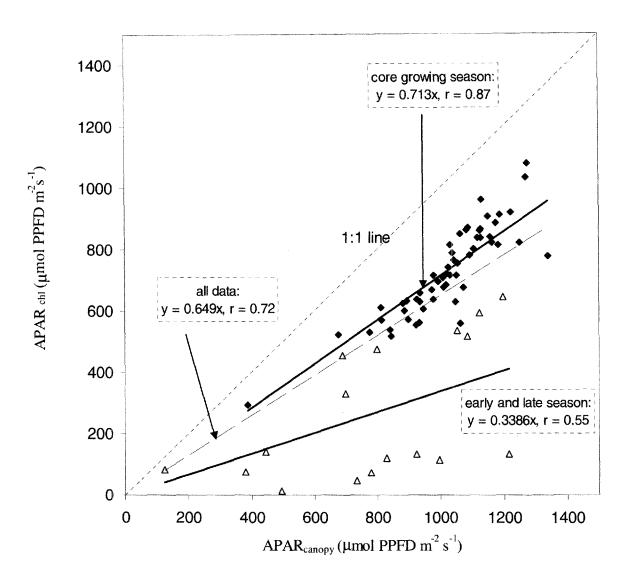


Fig. 2. The relationship of APAR_{chl} = [(90 min PPFD)*FAPAR_{chl}] to APAR_{canopy} = [(90 min PPFD)*FAPAR_{canopy}] for 2001-2005. Solid diamonds (\blacklozenge) indicate values retrieved during the core growing season (DOY:152-259; r = 0.87; APAR_{chl} = 0.713 *APAR_{canopy}). Early (DOY<152) and late (DOY>259) season values were combined and are indicated with open triangles (\triangle) (r = 0.55; APAR_{chl} = 0.339* APAR_{canopy}). For all values (dashed line), r = 0.72 and APAR_{chl} = 0.649*APAR_{canopy}.

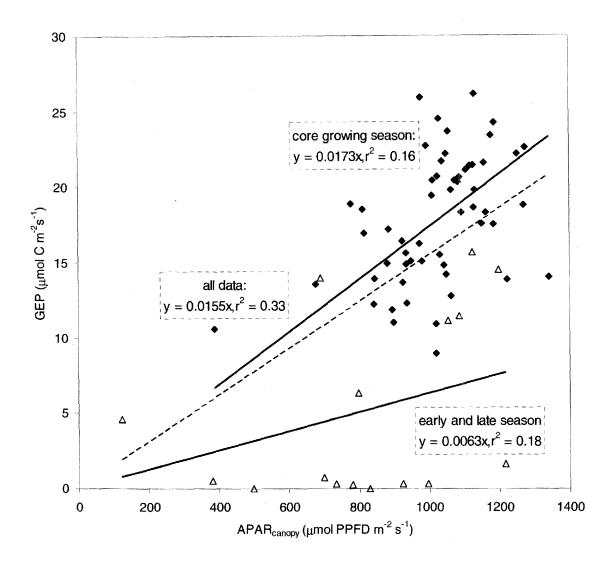


Fig. 3. The relationship of GEP to $APAR_{canopy} = [(90 \text{ min PPFD})*FAPAR_{canopy}]$ in 2001-2005. Solid diamonds (\blacklozenge) are data from during the core growing season (DOY 152-259,GEP=0.0173*APAR_{canopy},r^2=0.16). Open triangles (Δ) indicate values obtained before (DOY<152) or after (DOY>259)(GEP=0.0063*APAR_{canopy},r^2=0.18). The apparent LUE_{canopy}, the slope of the relationship (0.0155 µmol C mmol-1 PPFD, r² = 0.33), is similar to the core growing season value (0.0173) but is considerably higher than for days having low GEP in the spring and fall.

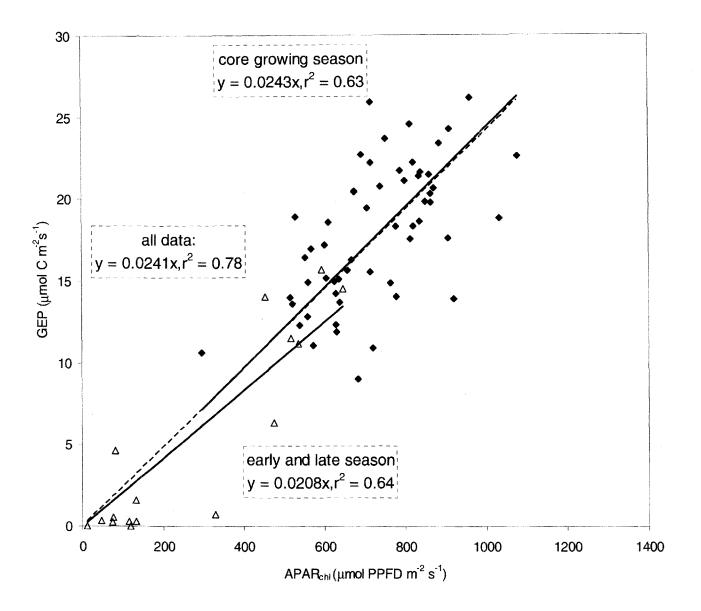


Fig. 4. The relationship of GEP to $APAR_{chl} = [(90 \text{ min PPFD})*FAPAR_{chl}]$ in 2001-2005. Solid diamonds (\blacklozenge) are data from during the core growing season (DOY 152-259, GEP = 0.0243*APAR_{chl}, r² = 0.63). Open triangles (Δ) indicate values obtained before (DOY<152) or after (DOY>259) (GEP = 0.0208*APAR_{chl}, r² = 0.64). The apparent LUE_{chl}, i.e., the slope of the relationship for all values, is 0.0241 µmol C µmol-1 PPFD (r² = 0.78).

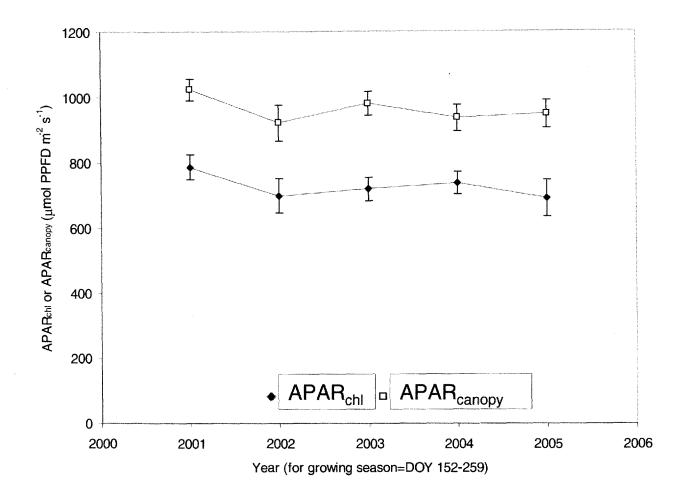


Fig. 5. Comparison of the annual means \pm SE for APAR_{chl} (\blacklozenge) and APAR_{canopy} (\Box) during the five-year period (2001-2005) for the growing season between DOY=152-259. APAR_{chl} was significantly lower than APAR_{canopy} in every year, averaging 235 µmol PPFD m⁻² s⁻¹ less than APAR_{canopy}. APAR_{tower} is not included in this figure because unlike APAR_{chl} and APAR_{canopy}, it was obtained under all sky conditions, i.e., both clear and cloudy.

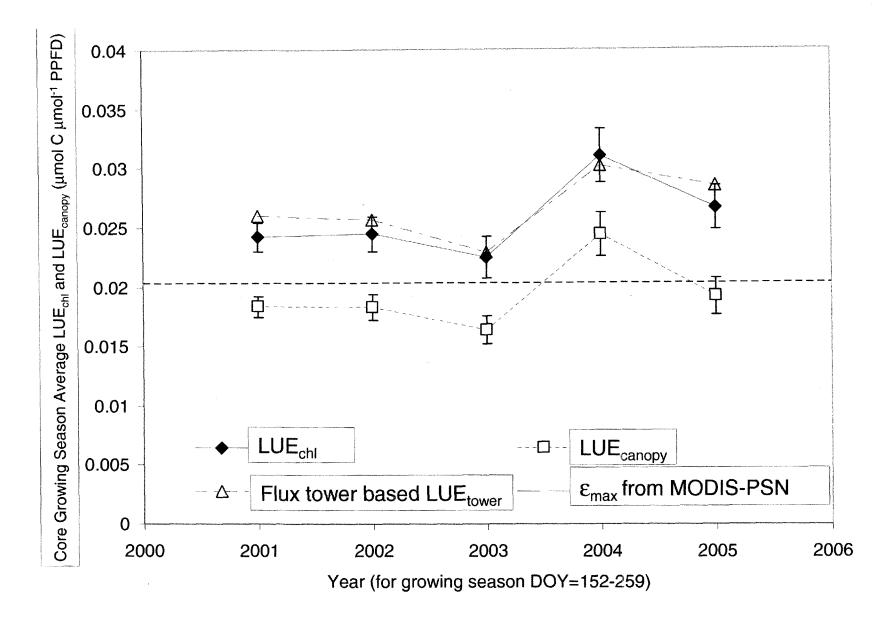


Fig. 6. Comparison of the annual means \pm SE for MODIS-derived LUE_{chl} (\blacklozenge) and LUE_{canopy} (\Box) during the five-year period (2001-2005) for the core growing season between DOY =152-259. LUE_{chl} was significantly higher than LUE_{canopy} in every year, averaging 0.007 µmol C µmol⁻¹ PPFD higher. Annual LUE_{tower} values (Krishnan *et al.*, 2006) (\triangle) agree well with our LUE_{chl}, falling within the SE range in 4 of 5 years. The maximum LUE (ε_{max}) used by the MODIS-PSN model is shown as a horizontal dashed line(---).

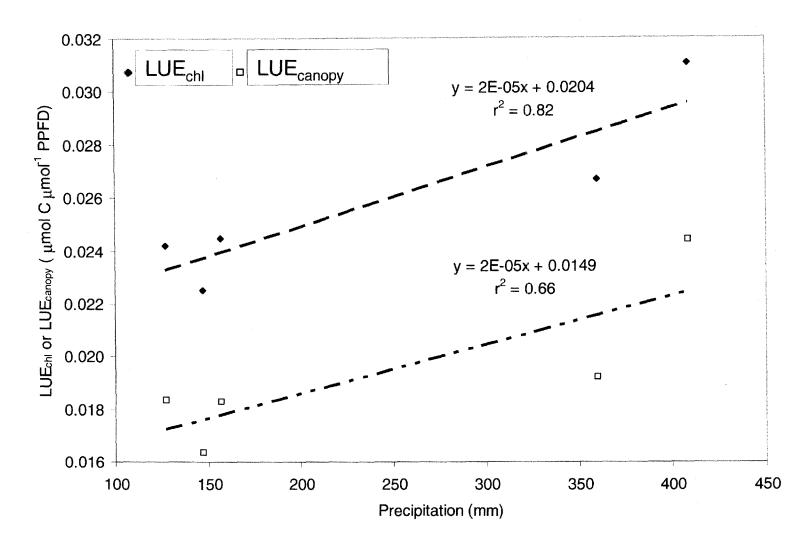


Fig. 7 Linear relationship between annual average LUE as a function of the cumulative precipitation for the core growing season (DOY=152–259) in 2001-2005.