

Table 1. Location and description of the sites used in this work

Site	Lat., °N	Long., °W	Description	Canopy N, % by mass	LAI, m ² m ⁻²	Age, years
Bartlett Experimental Forest, NH	44°03'	71°17'	Mixed northern hardwood	1.66	4.5	110–130
Donaldson Tract, FL	29°44'	82°13'	Slash pine	0.99	2.9	12
Duke Forest Deciduous, NC	35°58'	79°06'	Oak–hickory	1.85	5.6	90–100
Duke Forest Pine, NC	35°58'	79°05'	Loblolly pine	1.47	4.5	22
Harvard Forest, MA	42°32'	72°10'	Mixed deciduous	1.95	4.9	70
Howland, ME	45°12'	68°44'	Boreal evergreen	1.16	5.7	140
Hubbard Brook, NH*	43°57'	71°44'	Northern hardwoods	2.24	NA [†]	25
Morgan Monroe State Forest, IN	39°19'	86°24'	Mixed deciduous	2.06	4.9	60–90
Niwot Ridge, CO	40°01'	105°32'	Subalpine evergreen	0.93	4.0	100
Tremper Mount, NY*	42°05'	74°16'	Mixed deciduous	2.35	NA	60–80
Willow Creek, WI	45°48'	90°04'	Temperate deciduous	1.79	5.4	60–80
Wind River Experimental Forest, WA	45°49'	121°57'	Temperate evergreen	0.75	8.2	>500

All but two sites (marked with *) are part of the AmeriFlux research network.

[†]NA, not available.

partial least squares (PLS) regression with reflectance spectra from NASA's AVIRIS (Airborne Visible/InfraRed Imaging Spectrometer) and Hyperion (aboard the EO-1 satellite) instruments to yield mapped estimates of canopy %N for each site. Canopy %N was calculated for the core area of each tower's daytime footprint (16) by using predicted values from within a 250-m radius circle. A simple model of forest C exchange was combined with daytime CO₂ fluxes to produce estimates of instantaneous, canopy-level photosynthetic capacity (CA_{max}) for the landscape represented by each tower. The model included a function describing the response of photosynthesis to incident photosynthetically active radiation (PAR) and a term for ecosystem respiration, which allowed partitioning of the maximum potential assimilation rate from the tower measurements of net CO₂ exchange.

Results

Across all sites, CA_{max} was highly correlated with canopy %N (Fig. 1). The relationship was slightly stronger when CA_{max} was expressed per unit of canopy leaf mass ($r^2 = 0.76$) rather than per unit of ground area ($r^2 = 0.65$), although both trends were highly significant ($P < 0.01$). The mass-based relationship follows from well-established physiological relations (11), whereas expressing CA_{max} per unit of ground area relates more directly to tower-based measurements and is more relevant for models driven by remotely sensed data. The relationship between mass-based CA_{max} and canopy %N is nearly parallel to the leaf-level relationship between A_{max} and %N for woody plants worldwide (9; Fig. 1B). At any given %N, whole-canopy photosynthetic capacity is lower than that for individual leaves, which is expected given the considerable self-shading in multilayered canopies. Neither mass- nor area-based CA_{max} was correlated with LAI ($P > 0.50$ in both cases), likely because most of our study sites are closed canopy forests with relatively high LAI and near-complete interception of available light. Under these conditions, the photosynthetic potential per unit of leaf mass should be a better indicator of the efficiency with which intercepted light is used, rather than the total amount of foliage present. A significant positive relationship between light use efficiency (LUE) under full-sun conditions and canopy %N supports this idea ($P < 0.01$, $r^2 = 0.59$ [see supporting information (SI) Fig. S1 and SI Text]).

Whereas previous studies have demonstrated the ability to estimate canopy %N over small landscapes by using aircraft-based imaging spectroscopy (12), the scope of our dataset allowed us to examine relationships between %N and reflectance data obtained from global-scale satellite sensors. We

systematically degraded the spectral resolution of imaging spectrometer scenes from 10 nm to 20 nm, 30 nm and so on to 100 nm, and repeated the PLS analysis. This process yielded little loss of prediction accuracy, and the reflectance feature that emerged as the dominant source of variability with respect to %N at coarser resolutions was the height of the near-infrared (NIR) reflectance plateau (800–1,200 nm; Fig. 2). We also examined

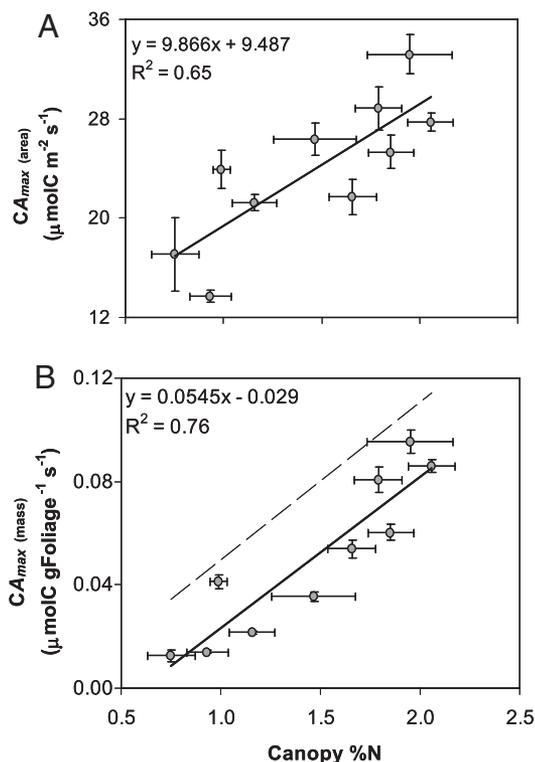


Fig. 1. Relationships among canopy-level CA_{max} and canopy N concentrations (%N) for the eddy covariance towers used in this work ($n = 10$). Area-based CA_{max} (A) was derived through inversion of a simple C flux model on tower net CO₂ exchange data. Mass-based CA_{max} (B) was derived by combining area-based values with canopy mass estimates for each site. Canopy %N values represent the average of estimates from imaging spectroscopy data, calculated from within a 250-m radius of each tower. The dashed line represents the leaf-level relationship between mass-based A_{max} and %N for woody plants worldwide (9). Uncertainty terms are standard errors of maximum likelihood parameter estimates (CA_{max}) and standard deviations of predicted values from within flux tower footprints (%N).

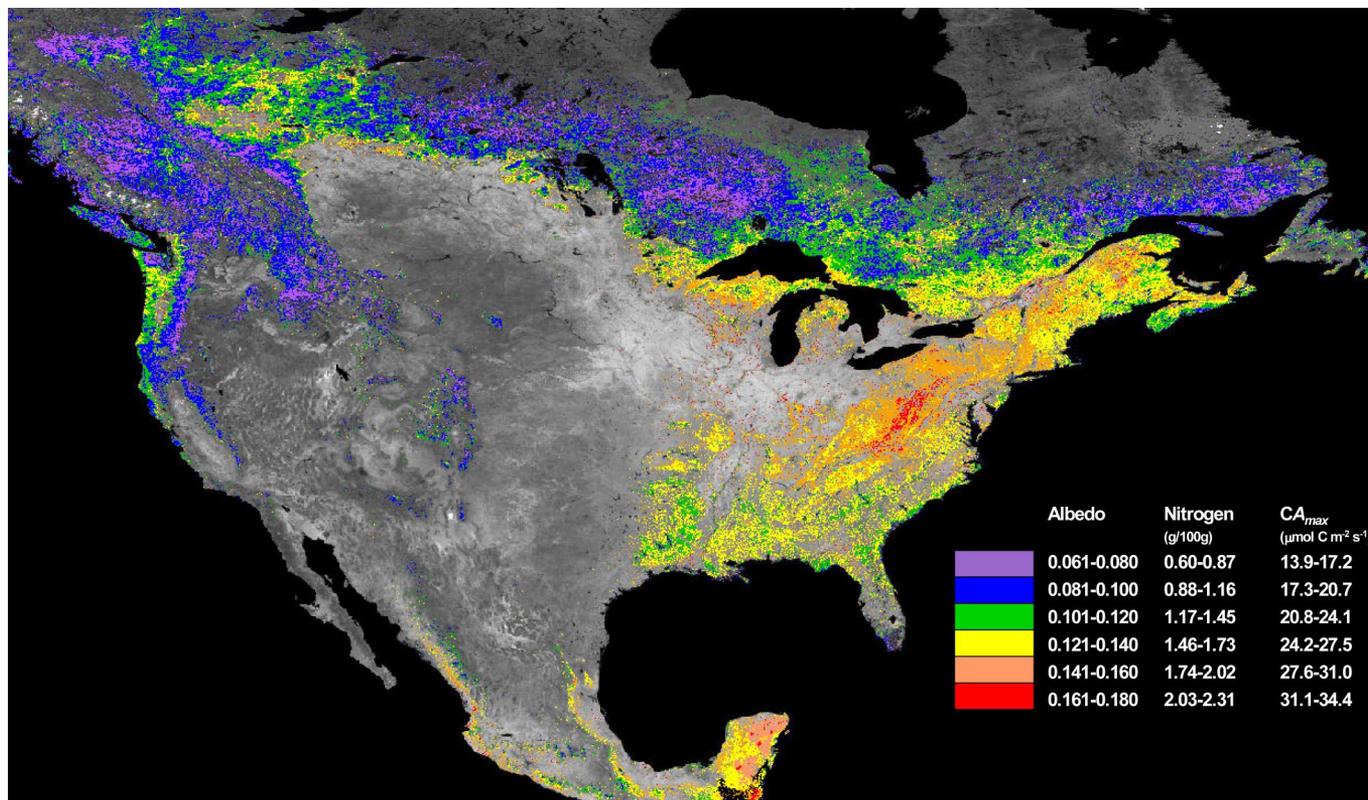


Fig. 4. Estimates of canopy %N and CA_{max} for forested areas of North America, along with the albedo values from which they were derived. Albedo data were from the midportion of the growing season and were converted to estimates of canopy %N and CA_{max} by using trends shown in Figs. 2C and 3B. Results were applied to forested grid cells only, identified by using the MOD44B continuous vegetation fields data product (41).

canopy temperate and boreal forests, NIR reflectance and albedo patterns are reliably correlated with canopy %N over a wide range of conditions with no underlying influence of variation in LAI. We therefore propose as an initial hypothesis that the dominant patterns of canopy %N and CA_{max} for North American forests can be estimated by combining the trends in Figs. 2C and 3B with a peak growing season coverage of MODIS shortwave albedo (Fig. 4). Although patterns related to dominant forest functional groups are evident (e.g., higher values in the eastern deciduous biome and lower values in northern evergreens), there is considerable variability within vegetation types, with the highest overall estimates in forests of the mid-Appalachian region. This region has been identified as having high foliar N concentrations relative to other temperate deciduous forests (20, 21), with some evidence that variation is linked to patterns of atmospheric N deposition (21). Additional field measurements can confirm or reject this hypothesis and determine whether or not similar trends can be applied over other regions and vegetation types.

Discussion

The strong linkages among canopy %N, shortwave albedo, and CA_{max} raise the possibility of an unrecognized feedback in the Earth's climate system involving the N cycle as a factor influencing surface energy exchange, in addition to its known effects on C assimilation. As an example, warming trends in boreal and arctic regions result in warmer soils, accelerated decomposition, and increased nutrient mineralization (22). Expected consequences include an increase in the N nutrition of plant canopies (23) or a shift in species composition to those requiring higher N supplies and having higher N concentrations in foliage. Although the effects of changes in major plant functional groups

have been included in coupled land surface–climate models (24), those associated with changes in N cycling or shifts in species composition within biomes have not. Based on our findings, these specific changes (i.e., increased N supply or altered composition, both resulting in higher canopy %N) should increase gross C assimilation [and presumably net C sequestration (4, 25)] and surface albedo, both representing negative feedbacks to warming. Similar effects can be expected from anthropogenic N deposition and widespread use of N fertilizers, which also serve to increase canopy %N (26). Conversely, leaf N concentrations in temperate forests are inversely related to mean annual temperature (27) and are driven down by rising CO_2 (6–8), suggesting that temperature and CO_2 increases could lead to reduced canopy albedo and a positive feedback to warming. Additional scenarios can be envisioned, but the overall implication is that alterations to plant N availability have important and unrecognized impacts on the climate system that warrant serious consideration (see *SI Text* for details on surface heating effects over the observed range of %N and albedo). This is particularly relevant in light of the degree to which human activities have altered the N cycle globally (28).

It would be premature to draw management implications from these results or to conclude that N fertilization, increased N pollutant emissions, or shifts from low-N to high-N tree species would help offset climate warming induced by greenhouse gas emissions. We also cannot assume that changes in C assimilation will always lead to a change in net C sequestration, given the potential for synergistic changes in ecosystem respiration. Moreover, interrelations among canopy %N, CA_{max} , and albedo need to be evaluated in additional ecosystem types, particularly in tropical forests, agricultural systems, and ecosystems with low LAI or sparse vegetation cover, and assessing the full outcome

of competing feedbacks will require new analyses using coupled climate–vegetation models. Finally, our analysis did not address N-induced changes in transpiration or potential consequences for cloud formation, which could have important effects on the net change in energy budgets (15) but were beyond the scope of our study. Despite these uncertainties, the observation of strong N–C–albedo linkages in forests add a new dimension to our understanding of the role played by ecosystems within the climate system and warrant consideration in climate projection efforts given the extent to which all of these variables are being altered by human activities.

Methods

Field Sampling. Canopy %N (g of N 100 g of dry leaf matter⁻¹) was sampled between 2001 and 2006 on 181 plots across the 12 research sites used in this study. At each site, plots were chosen from within a 5- × 5-km area to capture ecological variability in forest composition and condition that occurred over the local landscape. At each plot, all dominant and codominant species were identified, and between 2 and 5 trees per species were selected for green leaf collection from several heights in the canopy by using 12-gauge shotguns. Samples were collected to coincide with the peak of the growing season and with overflights by NASA AVIRIS or Hyperion instruments. A benchtop visible and near-infrared spectrophotometer (Foss NIRSystems 6500) was used to determine leaf-level foliar N concentrations of oven-dried, ground foliage according to methods described by Bolster *et al.* (29). The relative abundance of each species in the canopy was determined by using a point intercept sampling technique that quantifies the vertical distribution of leaf mass by species in the canopy (12). Plot-level canopy N concentrations (whole-canopy %N) were calculated as the mean of leaf-level foliar N concentrations among all species, weighted by the mass-based fraction of the canopy represented by each.

Canopy N Remote Sensing. Hyperspectral remote sensing, or imaging spectroscopy, is a form of optical remote sensing in which surface reflectance is examined in narrow and contiguous spectral regions that allow analysis of specific absorption features on a pixel-by-pixel basis. At each site, hyperspectral remote sensing scenes were obtained from NASA's AVIRIS or Hyperion [aboard the Earth Observing-1 (EO-1) satellite] sensors. AVIRIS measures upwelling radiance in 224 contiguous optical bands from 400 to 2,500 nm with a spectral resolution of 10 nm. AVIRIS was flown on an ER-2 aircraft at an altitude of 20,000 m, producing a spatial resolution of 17 m and a swath width of ≈10 km. The Hyperion instrument is part of the NASA EO-1 satellite, which was launched in November, 2000. The Hyperion detection capabilities are similar to those of AVIRIS (220 spectral bands ranging from 400 to 2,500 nm, with a 10-nm spectral resolution) but with a spatial resolution of 30 m. The swath width is 7.5 km, and data are typically collected in 7.5-km × 100-km images. Data from AVIRIS and Hyperion were converted from radiance (W m⁻² μm⁻¹ sr⁻¹) to surface reflectance by using the ACORN atmospheric correction program (ACORN 5.0).

The signal-to-noise ratio of AVIRIS is >600:1 over most of its spectral range, whereas the Hyperion signal-to-noise ratio does not exceed 200:1. For this reason, we used AVIRIS data where possible (9 sites), unless Hyperion was the sole option (3 sites). AVIRIS data yielded full-spectral coverage for all plots, whereas low signal-to-noise resulted in omission of some Hyperion bands. This had little effect on our analysis but may have reduced the accuracy of the Hyperion %N model and prevented full-spectrum average brightness values from being calculated for all plots and scenes.

Detailed methods for deriving canopy %N coverages from AVIRIS and Hyperion are given elsewhere (12). Relationships between plot-level spectra and canopy %N were examined by using PLS regression, in which the full-reflectance spectrum is collapsed into a smaller set of independent variables, or factors, with the measured %N data used directly during the spectral decomposition process. The accuracy of the resulting models was evaluated by using a standard 1-out cross-validation. The PLS models were then used to develop mapped estimates of canopy %N for a 5-km × 5-km landscape centered over each site. Canopy %N for each tower was calculated by using predicted values from within a 250-m-radius circle, corresponding to the core area of a typical tower's daytime footprint (16). Given uncertainties in the actual size of the tower footprints, we also calculated mean %N assuming footprint radii of 500, 750, and 1,000 m. Although the relationships in Fig. 1 generally grew weaker as the assumed footprint radius grew larger, the change was small, and the overall conclusions remained intact (e.g., from 250 m to 1,000 m, the *r*² of mass-based %N vs. *CA*_{max} declined from 0.72 to 0.58).

Flux Data and Derivation of *CA*_{max} and LUE Estimates. At each flux tower site, the eddy covariance method (30) was used to obtain continuous measurements of the surface–atmosphere exchange of CO₂, H₂O, and energy. Three years of flux measurements and ancillary meteorological data were used for each site. We used daytime measurements of the net ecosystem exchange of CO₂ (NEE, μmol m⁻² s⁻¹) to estimate 3 parameters of a simple “big leaf” Michaelis–Menten light response model of gross canopy photosynthesis driven by incident solar photosynthetic photon flux density (*I*, μmol m⁻² s⁻¹),

$$P_{\text{gross}} = \text{NEE} - R_{\text{eco}} = CA_{\text{sat}} \frac{I}{I + K_m} \quad \text{or}$$

$$\text{NEE} = P_{\text{gross}} + R_{\text{eco}} = CA_{\text{sat}} \frac{I}{I + K_m} + R_{\text{eco}} \quad [1]$$

where *CA*_{sat} is the theoretical, whole-canopy light-saturated rate of photosynthesis, *K*_m is the half-saturation value of *I* for the light response function, and *R*_{eco} is ecosystem respiration. Maximum likelihood parameter estimates were fit separately for each site, using 3 years of flux data and a 14-day moving window, following methods described by Press *et al.* (31) and by using flux uncertainty estimates presented by Richardson *et al.* (32). This approach to partitioning NEE to photosynthetic and respiratory component fluxes was recently evaluated and was found to yield results comparable to those of a range of other partitioning methods (33). Because of potential correlation between *CA*_{sat} and *K*_m parameters, which could result in unrealistically high values for the theoretical light-saturated rate of uptake, we instead focused subsequent analyses on *CA*_{max}, the maximum rate of assimilation when *I* = 2,000 μmol m⁻² s⁻¹, i.e., under reasonable “full-sun” conditions, and when other environmental variables are least limiting. Resulting values for *CA*_{sat} and *CA*_{max} were highly correlated (*r*² = 0.89, *P* < 0.0001), and use of either variable in the analyses produced the same overall conclusions. The *CA*_{max} values used for comparison with other variables in the study were means of the peak value achieved during each growing season of the 3-year record.

LAI and Maximum LUE. LAI values were obtained from a combination of published values and measurements taken during the time of our field sampling (34–38). All LAI data were from site-specific, midgrowing season measurements conducted with optically based methods, primarily by using the Li-Cor LAI-2000 (LI-COR Biosciences) instrument. Estimates based on leaf litterfall collections or alometry were also available for some sites. Although small discrepancies among LAI sampling methods have been noted (39), we used data collected from optically based methods to maintain consistency across sites.

Maximum canopy LUE (mol of C assimilated per mol of absorbed photons) was determined by dividing area-based *CA*_{max} by the quantity of absorbed photosynthetically active radiation (APAR) when *I* = 2,000 μmol m⁻² s⁻¹. APAR was estimated by using a simplified model of transmission and scattering of light in plant canopies, applied by using measured LAI values for each site (40).

MODIS Albedo and Surface Reflectance. MODIS reflectance and shortwave albedo data (collection 4, MOD43B4 and MOD43B, respectively) were obtained from the MODIS subset data server for each of the study sites (17). Individual image datasets are 16-day composites of atmospherically corrected surface reflectance, derived by using cloud-free images only. The reflectance product is nadir BRDF-adjusted reflectance (NBAR) and represents reflectance at the mean solar zenith angle. We used reflectance and albedo data from DOY 177 through 241 for years 2001–2006 to calculate mean growing season values that would be compatible with field measurements and hyperspectral imagery. Data completeness for this time period averaged 93% with gaps caused by cloudiness and other data quality issues. Data gaps were distributed randomly in time. Shortwave albedo is calculated across a spectral range of 300–5,000 nm and was available as black-sky, white-sky, and actual albedo (17). Data shown in Fig. 2C and 3 are actual albedo, although use of other albedo variables produced nearly identical results (e.g., black-sky values were within 2% of actual albedo, and a regression of the 2 across the study sites produced a relationship with *r*² = 0.997 and a slope and intercept not significantly different from 1 and 0, respectively).

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