

1 Quantifying the effects of harvesting on carbon fluxes and 2 stocks in northern temperate forests

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14 15 **Abstract**

16 Harvest disturbance has substantial impacts on forest carbon (C) fluxes and stocks. The
17 quantification of these effects is essential for better understanding of forest C dynamics and
18 informing forest management in the context of global change. We used a process-based forest
19 ecosystem model, PnET-CN, to evaluate how, and by what mechanisms, clear-cuts alter
20 ecosystem C fluxes, aboveground C stocks (AGC), and leaf area index (LAI) in northern
21 temperate forests. We compared C fluxes and stocks predicted by the model and estimated at two
22 chronosequences of eddy covariance flux sites for deciduous broadleaf forests (DBF) and
23 evergreen needleleaf forests (ENF) in the Upper Midwest region of northern Wisconsin and
24 Michigan, U.S.A. The average normalized root mean square error (NRMSE) and the Willmott
25 index of agreement (d) for carbon fluxes, LAI, and AGC in the two chronosequences were 20%
26 and 0.90, respectively. Simulated gross primary productivity (GPP) increased with stand age,
27 reaching a maximum (1200–1500 g C m⁻² yr⁻¹) at 11–30 years of age, and leveled off thereafter
28 (900–1000 g C m⁻² yr⁻¹). Simulated ecosystem respiration (ER) for both plant functional types

1 (PFTs) was initially as high as 700–1000 g C m⁻² yr⁻¹ in the first or second year after harvesting,
2 decreased with age (400–800 g C m⁻² yr⁻¹) before canopy closure at 10–25 years of age, and
3 increased to 800–900 g C m⁻² yr⁻¹ with stand development after canopy recovery. Simulated net
4 ecosystem productivity (NEP) for both PFTs was initially negative, with net C losses of 400–700
5 g C m⁻² yr⁻¹ for 6–17 years after clear-cuts, reaching the peak values of 400–600 g C m⁻² yr⁻¹ at
6 14–29 years of age, and eventually stabilizing in mature forests (>60 years old) with a weak C
7 sink (100–200 g C m⁻² yr⁻¹). The decline of NEP with age was caused by the relative flattening of
8 GPP and gradual increase of ER. ENF recovered more slowly from a net C source to a net sink,
9 and lost more C than DBF. This suggests that in general ENF may be slower to recover to full C
10 assimilation capacity after stand-replacing harvests, arising from slower development of
11 photosynthesis with stand age. Model results indicated that increased harvesting intensity would
12 delay the recovery of NEP after clear-cuts, but this had little effect on C dynamics during late
13 succession. Future modeling studies of disturbance effects will benefit from the incorporation of
14 forest population dynamics (e.g., regeneration and mortality) and relationships between age-
15 related model parameters and state variables (e.g., LAI) into the model.

16

17 **1 Introduction**

18 Disturbance has been widely recognized as a key factor influencing ecosystem structure
19 and function at decadal to century scales (Magnani et al., 2007; Williams et al., 2012; Kasischke
20 et al., 2013). Harvesting is an important anthropogenic disturbance shaping North American
21 forest landscapes. Approximately 61,000 km² of forests were affected by harvests every year
22 during the 2000s (Masek et al., 2011). Harvests alter forest age structure and the forest carbon
23 (C) cycle (Magnani et al., 2007; Pan et al., 2011; Williams et al., 2012; Zhou et al., 2013a).
24 Quantifying the legacies of harvest disturbances under the context of climate change is essential
25 for predicting forest C dynamics, informing climate policy-making, and improving forest
26 management. Here, we focus on assessment of an ecosystem model's ability to assess the C
27 cycle response to clear-cut harvesting.

28 Harvests transfer living biomass C to harvested wood C and litter C, resulting in
29 successional changes in C fluxes and stocks. Leaf biomass increases rapidly in secondary
30 succession and then typically stabilizes at a certain level that is determined by light, water,
31 nutrient availability, and forest type (Sprugel, 1985). Gross primary productivity (GPP) thus

1 increases gradually over time, reaches its maximum in middle age, declining slightly thereafter
2 in response to nutrient limitations and aging (Odum, 1969; Chapin et al., 2002; Tang et al.,
3 2014). The successional change in plant respiration (i.e., autotrophic respiration) after stand-
4 replacing harvesting is similar to that of GPP, although the C use efficiency, the ratio of net
5 primary productivity(NPP) to GPP (or NPP/GPP), generally decline with forest age (DeLucia et
6 al., 2007). As a result of these patterns, along with increases in C loss through woody litterfall,
7 living tree biomass C gradually increases following a typical logistic growth curve (Odum, 1969;
8 Sprugel, 1985).

9 Heterotrophic respiration following stand-replacing harvesting can be stimulated at the
10 beginning of stand development because the removal of trees alters the environmental conditions
11 (e.g., soil temperature, moisture, and nutrients) and possibly leads to increase in litter quantity
12 depending on harvest types (e.g., stem-only harvesting). Heterotrophic respiration is expected to
13 gradually decrease thereafter because the regrowing forest reduces net radiation, water, and
14 nutrient availability to the soil (Chapin et al., 2002) and the amount of decomposable soil organic
15 matter from the prior forest and harvest residue (e.g., litter, coarse woody debris, and soil organic
16 C) also gradually decreases. Over time, however, heterotrophic respiration could be enhanced
17 because of accumulation of woody debris and litter with stand development. This theorized
18 successional trajectory in ecosystem respiration (ER; the sum of autotrophic and heterotrophic
19 respiration) can also be influenced by harvest types and forest composition. Unlike GPP or NPP,
20 quantifying the trajectory of heterotrophic respiration (and consequently total ecosystem
21 respiration) with age is not as straightforward (Amiro et al., 2010). Observational studies have
22 shown that forest ecosystems generally become C sources (i.e., negative net ecosystem
23 productivity, NEP) immediately following stand-replacing harvests, approach the maximum
24 NEP as they mature, and then experience a gradual decline in NEP thereafter (e.g., Law et al.,
25 2003; Gough et al., 2007; Goulden et al., 2011), following the trajectories hypothesized by
26 Chapin et al. (2002).

27 The changes in C fluxes and stocks after harvesting have been examined in many forest
28 ecosystems using ecological measurements (e.g., eddy covariance or EC observations) from
29 chronosequences using a space-for-time substitution approach (e.g., Gough et al., 2007; Goulden
30 et al., 2011). The trajectory and amplitude of C fluxes and stocks vary with forest ecosystem
31 types (Amiro et al., 2010). For example, Noormets et al. (2007) reported that a young red pine

1 (*Pinus resinosa*) stand at 8 years of age was a net C sink ($313\pm 14 \text{ g C m}^{-2} \text{ yr}^{-1}$), but a young
2 hardwood site at age of 3 was a net C source ($-128\pm 17 \text{ g C m}^{-2} \text{ yr}^{-1}$) over the growing season in
3 northern Wisconsin, U.S.A. Young stands in northern Wisconsin may become net C sinks within
4 10-15 years after harvesting (Noormets et al., 2007). More rapid recovery after stand-replacing
5 harvesting (< 6 years) was found for temperate forests in northern Michigan (Gough et al., 2007).
6 These studies have produced a wealth of information on ecosystem C dynamics after stand-
7 replacing disturbances, and this information can be translated to more process-based and
8 quantitative understanding of disturbance effects on the C cycle using ecosystem models
9 (Goulden et al., 2011). Process models require evaluation on how source/sink transition and
10 long-term carbon flux dynamics respond to differences in vegetation type, harvest intensity, and
11 age since clearing.

12 Although using the chronosequence approach to evaluate the changes of ecological
13 processes with age after disturbances is attractive, this approach is often limited by the lack of
14 biological and climatic data (Yanai et al., 2003; Bond-Lamberty et al., 2006) and full
15 representation of stand development stages. Process-based ecosystem models provide a means of
16 quantifying the effects of disturbances on C dynamics under changing climate over various
17 spatial and temporal scales. Ecosystem models have been used to assess the effects of clear-cuts
18 and climate change on forest C dynamics at the stand/ecosystem (e.g., Bond-Lamberty et al.,
19 2006; Grant et al., 2009; Wang et al., 2012b) or regional scales (Desai et al., 2007; Dangal et al.,
20 2014). Moreover, ecosystem models can also be used to assess forest C dynamics under various
21 scenarios of climate change and harvesting regimes (e.g., Albani et al., 2006; Peckham et al.,
22 2012), since these models have been developed based on physiological, biogeochemical, and
23 ecological theories. However, few studies have used ecosystem models to examine the changes
24 of C fluxes and stocks with stand regrowth after stand-replacing disturbances for forest
25 chronosequences.

26 The objectives of this study were to evaluate the ability of an ecosystem model to capture
27 the trajectories of forest C dynamics after stand-replacing harvests for two northern temperate
28 plant functional types (PFTs: deciduous broadleaf forests, DBF; evergreen needleleaf forests,
29 ENF), to examine which processes influence successional trajectories in these ecosystems, and to
30 test the role of PFT on successional trajectory of C fluxes. We applied a process-based forest
31 ecosystem model, PnET-CN (Aber et al., 1997; Ollinger et al., 2002), to simulate the effects of

1 clear-cut on forest C dynamics, and evaluated the simulated C fluxes and stocks for both PFTs
2 using in-situ measurements (e.g., EC observations and aboveground biomass C, AGC). We
3 hypothesized that (1) both DBF and ENF will have similar successional patterns in C fluxes
4 (GPP, ER, and NEP) and aboveground biomass C stocks after stand-replacing harvests, but (2)
5 DBF will recover faster than ENF from a net C source to a net C sink and lose a smaller amount
6 of C (negative NEP) following a stand-replacing harvest.

7 **2 Methods**

8 **2.1 Study sites and field data**

9 Our study sites consist of 8 EC sites in the Upper Midwest region of northern Wisconsin
10 and Michigan (Chen et al., 2008; Table 1). The study area is characterized by a humid-
11 continental climate with hot summers and cold winters. The mean annual temperature is 4.4 °C
12 and the mean annual precipitation is 768.9 mm (as measured between 1981 and 2010 at Rest
13 Lake weather station, 46.12° N 89.87° W, <http://www.ncdc.noaa.gov>). The dominant soil type is
14 glacial sandy loam and loamy tills (Noormets et al., 2008). The region has been strongly
15 influenced by forest industry. Most forest stands are less than 100 years old in this region, having
16 regenerated following past harvesting (Amiro et al., 2010).

17 Our sites consist of four DBF sites (YHW, IHW, WIC, and UMBS) and four ENF sites
18 (YRP, YJP, IRP, and MRP). The four DBF sites range from 3 to 86 years in age and constitute a
19 chronosequence. Dominant tree species are bigtooth aspen (*Populus grandidentata*), trembling
20 aspen (*Populus tremuloides*), sugar maple (*Acer saccharum*), red maple (*Acer rubrum*), red oak
21 (*Quercus rubra*), basswood (*Tilia american*), and green ash (*Fraxinus pennsylvanica*). The four
22 ENF sites also represent a chronosequence with stand age ranging from 8 to 66 years. Red pine
23 and jack pine (*Pinus banksiana*) are the dominant tree species in the four ENF sites. At the two
24 chronosequences, most sites were initiated by stand-replacing harvests. We obtained monthly C
25 fluxes (observed NEP and its inferred data products GPP and ER) from AmeriFlux
26 (<http://public.ornl.gov/ameriflux/>) for the eight EC flux tower sites (Table 1). Harmonized level
27 4 data were used in this study. These flux data have been described and used in our previous
28 studies (e.g., Noormets et al., 2007; Chen et al., 2008; Desai et al., 2008; Xiao et al., 2011; Xiao
29 et al., 2014). We also obtained leaf area index (LAI) and AGC data from the literature for each
30 site (Table 1).

1 **2.2 Model description**

2 The PnET-CN model is a process-based forest ecosystem model designed to simulate C,
3 nitrogen (N), and water dynamics at daily to monthly time steps. PnET-CN is driven by climate
4 variables (temperature, precipitation, photosynthetically active radiation (PAR)), site variables
5 and atmospheric properties (soil moisture, disturbance history, wet and dry N deposition, and
6 atmospheric CO₂ concentration), and vegetation input parameters describing physiological and
7 structural plant traits (Aber and Driscoll, 1997; Aber et al., 1997; Ollinger et al., 2002). The
8 model has been applied and tested in the USA and Europe for simulating the effects of climate
9 variability, rising atmospheric CO₂, ozone pollution, and disturbance on ecosystem processes and
10 functions (e.g., Aber et al., 2002; Pan et al., 2009; Peters et al., 2013).

11 A characteristic feature of PnET-CN is its use of generalized leaf trait relationships to
12 simulate potential photosynthesis in a multilayered canopy (Aber and Federer, 1992). Actual
13 photosynthesis is then constrained by air temperature, vapor pressure deficit, and soil water
14 availability for simulating actual GPP. The effects of elevated CO₂ concentration on leaf
15 photosynthetic rates are calculated using constant ratios of leaf internal to ambient CO₂
16 concentration (C_i/C_a) (Ollinger et al., 2002). PnET-CN incorporates a total of seven C pools, five
17 of which are structural C pools (foliage, woods, fine roots, woody debris, and soil organic
18 matter) and two of which are non-structural C pools stored in woods and fine roots.
19 Photosynthetic production is allocated to each living plant component (i.e., foliage, woods, and
20 fine roots) and to growth and maintenance respiration. Living biomass is transferred to dead
21 woody biomass and/or to soil organic C through leaf, root and wood turnover, tree mortality, and
22 disturbance. The decomposition of coarse woody debris is a constant fraction of its C content.
23 The decomposition of soil organic C is calculated as a function of maximum decomposition rate
24 and effects of temperature and soil moisture.

25 PnET-CN includes a complete N cycle, and simulates N mineralization and nitrification,
26 plant N uptake, allocation, and leaching losses. N deposition is added to corresponding soil N
27 pools (NH₄ and NO₃). As with C pools, N is divided into five structural pools (foliage, woods,
28 fine roots, woody debris, and soil organic matter) and one non-structural N pool stored in the
29 trees. C and N cycles interact closely in the model. High leaf N concentration increases net
30 photosynthesis rate in the absence of water stress, thereby resulting in the high demand for non-

1 structural N in plant tissues (Aber et al., 1997). When plant non-structural N is low, plant N
2 uptake efficiency from available soil mineral N is increased in the model (Aber et al., 1997). In
3 addition, high C:N ratios in biomass, litter, and soil organic matter reduce net mineralization
4 rates.

5 The model also simulates key hydrological processes including rainfall interception,
6 evaporation, transpiration, surface runoff, and drainage at each time step. Rainfall interception is
7 treated as a constant fraction of precipitation. Transpiration is estimated based on water use
8 efficiency and plant demand via photosynthesis. Surface runoff is calculated as a constant
9 fraction of the difference between precipitation and evaporation. Drainage is estimated when
10 potential soil water exceeds soil water holding capacity.

11 Prescribed disturbance events can be simulated in the model through four parameters:
12 disturbance year, disturbance intensity, biomass removal fraction (live and dead), and the loss
13 rate of soil organic matter. In this study, when stand-replacing disturbance events occur, a
14 uniform PFT was assumed to be regenerated on-site. For the first year after clear-cuts, minimum
15 LAI of 0.1 was assumed to regulate maximum potential foliage mass that controls leaf
16 production. The photosynthetic production is transported to plant non-structural C pool where C
17 could be allocated to leaves, stems, and roots. There is, therefore, no need for initialization (e.g.,
18 stand density) after disturbances in the model. More details about the model structure and
19 processes have been described elsewhere (Aber et al., 1997; Ollinger et al., 2002).

20 **2.3 Model inputs**

21 The model inputs include temperature, precipitation, PAR, wet and dry N deposition,
22 atmospheric CO₂ concentrations, and disturbance history. The climate data used in all
23 simulations were derived from the Daymet database (Thornton et al., 2012). For each site,
24 monthly maximum temperature, minimum air temperature, and precipitation were calculated
25 from the daily Daymet data for the period 1981-2010. PAR ($\text{mol m}^{-2} \text{s}^{-1}$) was estimated from
26 solar radiation (RAD, $\text{MJ m}^{-2} \text{day}^{-1}$) using the empirical relationship ($\text{PAR} = 2.05 \text{ RAD}$) (Aber et
27 al., 1996). The data from 1981 through 2010 were repeated as needed to create the time series
28 from 1850 to 1980.

29 Annual rates of wet and dry N deposition were obtained from the United States
30 Environmental Protection Agency (EPA; <http://java.epa.gov/castnet/clearsession.do>). The N

1 deposition rates were measured at the Wellston station (44.22° N; 85.82° W) for the period 1994-
2 2011. We also obtained the N deposition rates in 1860 estimated by Galloway et al. (2004). For
3 each year prior to 1994, we used an exponential ramp function to estimate the annual deposition
4 rates by interpolating the historical (1860) and current N deposition rates. Monthly wet
5 deposition rates, needed for the model, were generated from annual wet N deposition through the
6 weighted coefficients (the ratio of monthly precipitation to total precipitation from March to
7 November). We assumed that there is no wet N deposition in the winter. The soil water holding
8 capacity in the rooting zone (100 cm) for each site was derived from the gridded multi-layer soil
9 characteristics dataset (STATSGO, Miller and White, 1998). For the period 1959-2010, we used
10 the CO₂ concentrations data from Mauna Loa. For the time period 1901–1958, we derived the
11 time series of the historical atmospheric CO₂ mixing ratio using a spline fit to the ice-core record
12 (Etheridge et al., 1996), as described by McGuire et al. (2001) and used by Xiao et al. (2009).
13 We used the CO₂ concentration in 1901 for the simulation period prior to 1901 and spin up.

14 For each site, we prescribed the disturbance events using the site disturbance history
15 (Table 1). For each stand-replacing harvest, stand mortality was assumed to be 100%. The
16 merchantable wood removal (biomass removal out of the ecosystem) fraction was assumed to be
17 0.8 in this study. The soil removal fraction was assumed to be zero, given that the content of soil
18 organic C might not be considerably affected by harvesting (Johnson and Curtis, 2001; Yanai et
19 al., 2003). We also conducted a sensitivity analysis to these assumptions as described below in
20 section 2.4.

21 **2.4 Parameterization, initialization, validation, and sensitivity analysis**

22 PnET-CN has been parameterized and tested for temperate DBF (Aber et al., 1997;
23 Ollinger et al., 2002; Peters et al., 2013), temperate ENF (Aber et al., 1997; Peters et al., 2013),
24 and mixed forests (Aber et al., 1997) for forest productivity, net N mineralization, and foliar N
25 concentrations. The parameter values used in this study are given in supplement Table S1. To
26 apply the model to the transient simulation period (1860-2010), a 200-year spin up run was
27 conducted to ensure that the equilibrium (Δ NEP < 10 g m⁻² month⁻¹ and Δ soil organic C < 1%)
28 was reached for each chronosequence site. Thirty year climate normals (1981-2010), pre-
29 industry N deposition rates, and historical CO₂ concentrations were used for the spin up runs.

1 To examine stand-replacing harvest legacies, we conducted all simulations using the site
 2 disturbance history (Table 1), vegetation parameters (Supplement Table S1), climate, N
 3 deposition, and atmospheric CO₂ for each of the chronosequence sites. The model simulations
 4 were evaluated against C fluxes (GPP, ER, and NEP), AGC, and LAI data collected at the EC
 5 flux sites. We used two statistical measures to evaluate the overall model performance: the
 6 normalized root mean square error (NRMSE) and the Willmott index of agreement. The NRMSE
 7 (Eq. (1)) was used to assess the difference between predicted (*P*) and observed (*O*) variables, and
 8 can be expressed as:

$$9 \quad \mathbf{NRMSE} = (\mathbf{O}_{max} - \mathbf{O}_{min})^{-1} \left[\frac{\sum_{i=1}^n (P_i - O_i)^2}{n} \right]^{0.5} \times \mathbf{100\%} \quad (1)$$

10 where *O*_{max} and *O*_{min} are the maximum and minimum observed values, respectively; *i* is the *i*th
 11 observation; and *n* is the total number of observations. A value close to 0 indicates perfect
 12 agreement and a value of 100% suggests poor agreement. The Willmott index of agreement (*d*) is
 13 an indicator of modeling efficiency and is expressed as:

$$14 \quad d = 1 - \left[\frac{\sum_{i=1}^n (P_i - O_i)^2}{\sum_{i=1}^n (|P_i| + |O_i|)^2} \right] \quad (2)$$

15 A value of 1 indicates perfect agreement and a value near 0 indicates weak agreement (Willmott,
 16 1982).

17 The sensitivity of ecosystem C dynamics to changes in harvesting practices during the
 18 secondary succession was assessed using sensitivity analysis. The model was run at WIC and
 19 MRP for 100 years after scenario harvests in 1910 using the same climate data sequence.
 20 Sensitivity scenarios involved applying the stand mortality (80% and 60%, compare to 100% in
 21 the model test), soil organic matter loss (20% and 40%, compare to zero in the model test) to
 22 assess the effects of different harvest intensity and soil organic matter loss on C dynamics. We
 23 also tested the model sensitivity to CO₂ fertilization for evaluating potential climate change
 24 effects.

1 **3 Results**

2 **3.1 Evaluation of modeled carbon fluxes and stocks**

3 Simulated C fluxes were generally consistent with EC derived C fluxes for both DBF and
4 ENF sites (Figs. 1 and 2). The NRMSE between simulated and tower fluxes (GPP, ER, and NEP)
5 were between 10% and 21% (Table 2). The Willmott index of agreement between simulated and
6 tower C fluxes for both PFTs ranged from 0.91 to 0.94 with the exception of NEP ($d= 0.73$, $n=$
7 235). The model underestimated GPP for the DBF sites and predicted ER fairly well for all DBF
8 sites, except for the intermediate-aged hardwood site, IHW. As a result, the model
9 underestimated NEP for most DBF sites. For IHW, the model substantially underestimated both
10 GPP and ER but predicted NEP fairly well. For the ENF sites, the model underestimated GPP.
11 The model predicted ER fairly well for YRP (8 years old), YJP (15-16 years old) and IRP (23
12 years old), but overestimated ER for the older MRP sites. Thus, the model underestimated NEP
13 for the ENF sites.

14 The simulated and observed stand characteristics (LAI and AGC) showed good
15 agreement (Table 2 and Fig. 3). The model slightly underestimated LAI for the young forest sites,
16 and overestimated LAI for the mature forest sites. Generally, the model overestimated AGC for
17 the mature forest sites. The NRMSE was 28% for AGC and 31% for LAI. The Willmott index of
18 agreement was 0.95 and 0.96 for AGC and LAI, respectively. Overall, the model evaluation
19 metrics indicated that the model performed better in the DBF sites than in the ENF sites.

20 **3.2 Effects of clear-cutting on carbon fluxes and stocks**

21 PnET-CN generally captured the changes of C fluxes following the clear-cuts for each
22 chronosequence site (Fig. 4). The predicted annual GPP generally increased with time since
23 disturbance and approached the peak values ($1200\text{--}1500\text{ g C m}^{-2}\text{ yr}^{-1}$) between 11 and 26 years
24 of age and between 29 and 30 years of age for the DBF (IHW, WIC, and UMBS) and the ENF
25 (IRP and MRP) sites, respectively; thereafter, the forest stands reached maturity and GPP
26 became relatively stable with mean values of $940\text{--}1000\text{ g C m}^{-2}\text{ yr}^{-1}$.

27 Predicted annual ER was initially as high as $860\text{--}1030$ and $710\text{--}860\text{ g C m}^{-2}\text{ yr}^{-1}$ within
28 the first two years for the DBF and the ENF sites, respectively. During canopy recovery,
29 predicted ER generally decreased to $620\text{--}780\text{ g C m}^{-2}\text{ yr}^{-1}$ between 10 and 25 years of age for the

1 DBF sites and to 360–380 g C m⁻² yr⁻¹ between 14 and 17 years of age for the ENF sites (Fig. 4).
2 For forest age older than 60 years, the predicted annual ER for both PFTs showed a relatively flat
3 pattern, contrary to theoretical expectations, arising from the little change of both autotrophic
4 and heterotrophic respiration with age (Supplement Fig. S1). Average annual ER for mature
5 forests was 810–880 and 780 g C m⁻² yr⁻¹ for the DBF sites (WIC and UMBS) and the ENF
6 (MRP) site, respectively.

7 As expected, the ratio of annual GPP to annual ER (GPP:ER) simulated by PnET-CN
8 was low during the early years after clear-cutting for both DBF and ENF (Fig. 5). Within ~6
9 years for the DBF sites and ~17 years for the ENF sites, the GPP:ER ratio gradually increased
10 and its average value became larger than 1 (NEP>0). The simulated peak GPP:ER ratio for DBF
11 (1.6) occurred at 18 years after stand-replacing harvests, and the simulated peak ratio for ENF
12 was 1.8 at 26 years. After those peaks, the ratio became relatively stable, with the mean values of
13 1.1 and 1.2 for mature DBF and mature ENF, respectively.

14 The model predicted negative NEP (C source) for the first 6 and 17 years after stand-
15 replacing harvests for the DBF and the ENF, respectively (Fig. 4). The simulated peak annual net
16 C loss occurred in the first or second year after clear-cutting. The average C loss was 530–710 g
17 C m⁻² yr⁻¹ for the DBF sites and 380–400 g C m⁻² yr⁻¹ for the ENF sites. The total C loss was 3.2–
18 4.3 and 6.4–6.9 kg C m⁻² for the DBF and the ENF sites, respectively. The maximum net C gain
19 was 387–433 g C m⁻² yr⁻¹ at 14–26 years of age for the DBF sites (WIC and UMBS) and was
20 567–602 g C m⁻² yr⁻¹ at 29 years of age for the ENF sites (IRP and MRP). Simulated annual NEP
21 decreased thereafter and became as low as 120–180 g C m⁻² yr⁻¹ after 17–31 years for the DBF
22 sites and 170 g C m⁻² yr⁻¹ after 44 years for the ENF sites.

23 LAI fully recovered within 10–15 years after disturbance for the DBF sites and within 40
24 years of age for the ENF sites (Fig. 6). The recovery of LAI led to the gradual increase in GPP
25 and the subsequent increase in AGC (Fig. 7). In general, AGC recovered much more slowly than
26 C fluxes and LAI. The changes of simulated AGC followed the logistic growth curve with slow
27 accumulation in the early years, fast accumulation in the middle age, and slow accumulation
28 afterwards. The predicted LAI and AGC generally fell within the range of observed values
29 across two chronosequences (Figs. 3, 6, 7). For mature forests (>60 years of age) in 2010, the

1 DBF sites generally stored more C in aboveground biomass than the ENF sites (10–12 vs. 8.5 kg
2 C m⁻²; Fig. 7).

3 **3.3 Sensitivity analysis**

4 Harvest intensity had little effect on long-term C dynamics for both PFTs, but it had
5 sizeable effects during early succession (Fig. 8). Increasing harvest intensity delayed GPP (Fig. 8
6 a and f) and LAI (Fig. 8 d and i) rises and led to lower reduction in ER (Fig. 8 b and g), resulting
7 in later rising NEP (Fig. 8 c and g). High harvest intensity (e.g., 100% removal of living trees)
8 also directly reduced living tree AGC (Fig. 8 e and i). By reducing harvest intensity parameter to
9 80% and 60% from 100% used in the original model, average annual NEP over 100 years for
10 DBF decreased by 104 and 88 g C m⁻² yr⁻¹, respectively. The increased remaining tree biomass
11 resulted in an increase in AGC about 12% and 16%, respectively, after a 100-yr harvest cycle.
12 For ENF average annual NEP decreased about 1% and AGC decreased nearly 6% for both
13 reduced harvest intensity scenarios. Increasing the soil removal fraction parameter resulted in
14 lower GPP and ER along succession and lower NEP in middle succession for both DBF and ENF
15 (Fig. S2 a-c and f-h). Greater soil removal fraction enhanced the leaf biomass of DBF in middle
16 and late succession (Fig. S2 d), but restricted the leaf biomass of ENF in late succession (Fig. S2
17 i). Increasing the soil removal fraction parameter (20% and 40% removal of soil organic matter)
18 strongly reduced living AGC (16% and 39%, respectively) for DBF (Fig. S2 e) but slightly
19 decrease living AGC (up to 5%) for ENF (Fig. S2 j). There were insignificant effects of CO₂
20 fertilization on carbon dynamics for both DBF and ENF in our sensitivity analysis (Fig. S3).

21 **4 Discussion**

22 **4.1 Carbon fluxes and stocks following clear-cutting**

23 PnET-CN generally simulated the expected post-harvest trajectories in C fluxes (GPP,
24 ER, and NEP) and stock (LAI and AGC). The model was unable to simulate high GPP rates
25 estimated by the EC technique in mature forests regardless of vegetation type, suggesting that
26 there is room for improvement in model simulation of secondary succession.

27 Our simulations showed that LAI increased rapidly first and then stabilized during the
28 following development stages, because the model estimates foliage growth through the
29 parameter of maximum relative growth rate (Table S1) with the restriction of current foliage

1 biomass and resource availability. This modeled response is consistent with the previous finding
2 that foliage biomass increased rapidly after disturbance and then stabilized (Sprugel 1985). Our
3 chronosequence-based results are generally consistent with previous results. For example,
4 Goulden et al. (2011) observed that LAI along a chronosequence of boreal forest stands
5 increased rapidly from $0.3 \text{ m}^2 \text{ m}^{-2}$ one year after fire, and then generally leveled off at $5.3\text{--}7.2 \text{ m}^2$
6 m^{-2} from 23 to 154 years after stand-replacing crown fire. A modeling study based on a modified
7 version of Biome-BGC (Bond-Lamberty et al., 2005) also showed a similar successional change
8 in LAI for boreal DBF and ENF.

9 The simulated successional change in annual GPP for both PFTs generally followed the
10 trajectory hypothesized by Odum (1969). However, despite a slight decrease in GPP
11 hypothesized in Odum's trajectories, our simulations show a relatively flat GPP in mature forests
12 (Figs. 4 and 10). In the model, GPP tracks LAI in the absence of significant changes in light,
13 water or nutrient stress. As LAI stabilizes in mature forests, GPP also stabilizes. Our results are
14 consistent with previous studies showing relatively flat pattern in GPP after 20 years following
15 harvests in temperate pine forests in Florida (Clark et al., 2004), northern temperate DBF in
16 Wisconsin (Desai et al., 2008), and boreal jack pine forests in Saskatchewan (Zha et al., 2009).
17 Furthermore, Humphreys et al. (2006) reported continuous increases of GPP with increasing
18 forest age for temperate rainforests using three different stands at different stages of development
19 (2, 14, and 53 years of age) following clear-cuts in British Columbia, Canada. However, northern
20 temperate ENF showed a small difference in GPP between young and mature sites (Noormets et
21 al., 2007; Desai et al., 2008). Desai et al. (2005) found that a nearby old-growth mixed forest had
22 slightly lower GPP and significantly higher ER than nearby DBF sites. Site-to-site variations in
23 species and soil fertility could result in variations in the successional trajectory of GPP after
24 clear-cuts such that the observed trajectories may deviate from hypothesized or modeled
25 trajectories. In addition, our chronosequences lack old-growth sites and do not encompass the
26 full range of forest development stages, which limits the representativeness of the C flux and
27 stock trajectories derived from chronosequence studies based on EC or other ecological
28 observations (e.g., Clark et al., 2004; Humphreys et al., 2006; Noormets et al., 2007).

29 We found that annual ER for secondary temperate forests declined slightly in the first ten
30 years because of low autotrophic respiration at first after the removal of trees. Amiro et al.
31 (2010) reported that ER was reduced in the very first year following harvests for a number of EC

1 flux sites over North America. Previous field studies showed that ER following clear-cuts
2 increased with forest age (e.g., Humphreys et al., 2006; Zha et al., 2009), partly supporting our
3 results that ER slightly increased after the short decline period (10-25 years of age) in northern
4 temperate forests until the stands reached maturity. Martin and Bolstad (2005) showed that
5 chamber-based soil respiration in DBF of northern Wisconsin ranged from 857–1143 g C m⁻² yr⁻¹
6 in 1998 and 1013–1357 g C m⁻² yr⁻¹ in 1999, which is higher than tower ER (825±133 g C m⁻² yr⁻¹
7¹, WIC) from 1999-2006 in the same region. Tang et al. (2009) reported that growing-season soil
8 respiration in 2005 was 690 g C m⁻² in a mature DBF near WIC tower site, which is in the range
9 of our simulations (Supplement Fig. S1). The model underestimated GPP, but estimated ER well
10 for mature DBF sites, indicating that the model likely overestimated autotrophic respiration. EC
11 derived ER were usually lower than chamber-based estimate at the WIC site due to uncertainties
12 induced by measurement methods, decoupling of surface and canopy fluxes at night, and spatial
13 scaling (Bolstad et al., 2004; Cook et al., 2004). For ENF, the model overestimated ER for the
14 mature site because of overestimated soil decomposition rate. Our simulations also show that
15 DBF had slightly higher soil respiration rate than ENF (Supplement Fig. S1), which is consistent
16 with the finding that chamber-based soil respiration was slightly higher for DBF than for ENF in
17 Wisconsin (Euskirchen et al., 2003). The changes of ER in secondary forests after clear-cutting
18 differ among sites because of different site conditions (e.g., quantity and quality of soil organic C
19 and litter C) and harvesting types (e.g., Tang et al., 2009).

20 The trajectory of our simulated GPP:ER ratio is similar to the curve derived by Amiro et
21 al. (2010) using EC observations and forest age from fire and harvest chronosequences across
22 North America ($GPP:ER = 1.23*[1-\exp(-0.224*AGE)]$). Our simulated ratios are within the
23 observed range of 0.9–1.6 for the DBF sites (Fig. 5a), although the model underestimated the
24 observed ratios for mature sites. Growing season GPP:ER ratios are typically higher than the
25 annual ratios because winter soil organic C decomposition is important to annual C balance
26 (Aanderud et al., 2013). However, the simulated ratios for the ENF sites are much lower than
27 tower-derived growth season ratios (1.9–4.7, Fig. 5b), and close to the annual range of 1.6–2.2
28 estimated by Desai et al. (2008). The standard gap-filling methods for the EC flux data may lead
29 to the overestimation of net ecosystem exchange due to the lack of winter C flux observations for
30 the ENF sites and two of the DBF sites (YHW and IHW).

1 Our simulated successional dynamics of NEP following clear-cuts generally supported
2 the trajectories of Chapin et al. (2002), but for different reasons. The hypothesized trajectories
3 show declining GPP and relatively flat ER with time. Our simulated decline in NEP resulted
4 from relatively flat GPP and growing ER with stand development (Figs. 4 and 8). This has been
5 observed for northern temperate hardwood chronosequence sites (Desai et al., 2008), northern
6 temperate pine forests (Peichl et al. 2010), and boreal DBF forests (e.g., Goulden et al., 2011). A
7 recent North American Carbon Program (NACP) synthesis study showed similar changes in NEP
8 after either stand-replacing fire or harvest based on EC chronosequence measurements across
9 North America (Amiro et al., 2010).

10 Chapin et al. (2002) hypothesized that heterotrophic respiration is initially high following
11 disturbance, declines in early succession, rises in middle succession, and declines thereafter,
12 while NPP reaches a peak in middle age and declines in old stands. The simulated successional
13 trajectories in heterotrophic respiration were supported by hypothesized pattern change, although
14 our simulated NPP did not decline in mature stands (Supplement Fig. S1). Previous studies also
15 support our simulated trajectory in heterotrophic respiration. For example, Pregitzer and
16 Euskirchen (2004) reported that heterotrophic respiration was high (mean value of 970 g C m^{-2}
17 yr^{-1}) in young temperate forests, declined with age in middle succession, and increased with time
18 for mature forests. A decline of NPP with age was not predicted in this study. Recent studies
19 found that a decline of NPP with age is not “universal” (Ollinger and Smith, 2005; He et al.,
20 2012). Validation of the simulated NPP was not possible in this study due to the lack of NPP
21 measurements across all sites. Our simulated heterotrophic respiration for mature DBF is close to
22 the observation of $502 \pm 86 \text{ g C m}^{-2} \text{ yr}^{-1}$ in a mature DBF near UMBS tower site between 1999
23 and 2003 (Gough et al., 2008). However, ENF chronosequence sites in this study show that NEP
24 continually increased with age because of relatively flat and low ER ($340 \pm 96 \text{ g C m}^{-2} \text{ yr}^{-1}$) and
25 increasing GPP.

26 Although our model underestimated NEP and GPP for both the DBF and ENF sites in the
27 Upper Midwest region (Figs.1 and 2), our predicted NEP was comparable to estimates from
28 other studies in similar regions. For example, our predicted maximum NEP for the ENF sites
29 ($567\text{--}602 \text{ g C m}^{-2} \text{ yr}^{-1}$, 29 years of age) was slightly lower than the estimates ($690 \text{ g C m}^{-2} \text{ yr}^{-1}$,
30 15–20 years of age) for afforested white pine (*Pinus strobus*) forests in Ontario (Coursolle et al.,
31 2012). For a northern temperate forest chronosequence study in northern Michigan, NEP higher

1 than 200 g C m⁻² yr⁻¹ in young DBF forests could be derived from the reference forest (153± 115
2 g C m⁻² yr⁻¹, 85 years of age) (Gough et al., 2007), suggesting that our predictions (390–430 g C
3 m⁻² yr⁻¹, 14–26 years of age) for the DBF sites could be in the reasonable range. Furthermore, our
4 predicted mean annual NEP (123–177 g C m⁻² yr⁻¹) for mature DBF sites (>60 years) was close
5 to estimates for other northern DBF, including a northern hardwood forest of central
6 Massachusetts (200±40 g C m⁻² yr⁻¹, Barford et al., 2001) and four eastern North American DBF
7 (167–236 g C m⁻² yr⁻¹ Curtis et al., 2002).

8 We found that the simulated AGC during forest regrowth gradually increased following
9 the typical logistic growth curve (Sprugel, 1985). In the model, low NPP in the early stages
10 results in slow AGC accumulation. Once the amount of NPP approximately equals annual dead
11 biomass C that is largely controlled by the wood turnover rate, the trajectory of AGC reaches a
12 plateau. Previous chronosequence studies showed that AGC increased with increasing age (e.g.,
13 Peichl and Arain, 2006; Goulden et al., 2011; Powers et al., 2012). Powers et al. (2012) reported
14 that AGC increased rapidly with age in young red pine stands across a chronosequence in
15 northern Minnesota, USA. The representativeness and generalization of these findings were
16 limited by the small number of young stands (Powers et al., 2012).

17 Sensitivity analysis shows that more intensive harvests could have larger and longer
18 impacts on successional trajectories of C dynamics in early succession for both DBF and ENF.
19 Fewer flux tower based studies have investigated the effects of harvest intensity on forest C
20 fluxes (e.g., GPP, ER, and NEP) because of the high establishment cost of EC systems.
21 Nevertheless, some modeling studies have provided insights into how forest C fluxes and stocks
22 are affected by harvest intensity. Our findings are supported by previous modeling studies. For
23 example, a recent modeling study of temperate forests reported that more intensive harvests
24 increased the recovery time of NPP for ENF and DBF in Minnesota and Wisconsin, USA (Peters
25 et al. 2013). In the boreal forest of central Canada, less intensive harvest and longer rotation
26 length might increase total C sink up to 40% (Peng et al. 2002), although recent studies indicate
27 that longer rotation length could not necessarily increase C sequestration under changing climate
28 conditions (Wang et al., 2012b; Wang et al., 2013). A recent synthesis study on the effects of
29 partial cutting on forest carbon stocks found that partial cutting has no significant effects on litter
30 C and soil organic C, although more intensive cutting can reduce more AGC (Zhou et al.,
31 2013b). This synthesis study is not able to determine the recovery duration due to the lack of

1 long-term observations. If harvesting operations largely reduce soil organic matter, C fluxes
2 (e.g., GPP, NPP, ER, and NEP) and living AGC are reduced for both PFTs. Consistent with this,
3 Peters et al. (2013) showed that simulated NPP could not recover to pre-harvesting levels due to
4 greater removal of soil organic matter. Therefore, our model results suggest management
5 practices should aim to decrease soil disturbance caused by harvest operations.

6 **4.2 Differences between deciduous broadleaf and evergreen needleleaf forests**

7 We found that DBF may reach a peak in LAI and GPP faster than ENF after clear-
8 cutting, showing clear differences in pattern of ecosystem development between the DBF and
9 ENF sites. More rapid recovery of LAI and GPP for DBF sites lead to sooner recovery of NEP
10 and AGC regardless of harvest intensity, supporting our second hypothesis. The foliage related
11 parameters such as FolRelGroMax and AmaxB mainly govern the differences in successional
12 trajectories between the two PFTs (Supplement Table S1). DBF is assumed to have more
13 productive foliage than ENF, and more photosynthetic production then can lead to more foliage
14 production. With this positive feedback in the model, GPP, NEP, and AGC of the DBF sites
15 recover more rapidly than those of the ENF sites. Our findings are consistent with the
16 chronosequence studies showing that the temperate DBF in northern Michigan rapidly became a
17 net C sink after six years following disturbances (Gough et al., 2007) and that ENF stands in
18 northern Wisconsin became net C sinks within 10-15 years after harvesting (Noormets et al.,
19 2007). Through the analysis of the Forest Inventory and Analysis (FIA) data, Williams et al.
20 (2012) suggested that faster growth in AGC at high productivity sites caused higher C fluxes and
21 stocks. Our findings are also consistent with a recent modeling study suggesting that temperate
22 DBF switches to positive NEP faster than temperate ENF after clear-cuts, and DBF has a higher
23 peak in NEP compared to ENF (Peckham et al., 2012). A modeling study conducted in boreal
24 forests also reported that low productive boreal ENF needed 1-3 more years to attain a positive
25 NEP than boreal DBF after clear-cuts in Saskatchewan, Canada (Wang et al., 2012b). These
26 observed and modeled successional changes further indicate that DBF tend to have higher
27 photosynthetic capacity than ENF in the early stage of stand development following stand-
28 replacing harvests.

29 The sensitivity analysis suggests that more productive forests could be more strongly
30 affected by greater soil removal fraction, as soil removal reduces soil organic matter thereby

1 resulting in relatively low N mineralization in the model. Peters et al (2013) showed that NPP
2 was more strongly reduced for Aspen than for jack pine in their simulations. However,
3 productive lodgepole pine (*Pinus contorta* Dougl. ex Loud.) could maintain high productivity at
4 12 years after harvest disturbance regardless of soil organic carbon removal and soil compaction
5 treatments because beneficial ectomycorrhizal fungi associated with lodgepole pine could help
6 access N from organic matter; while hybrid white spruce (*Picea glauca* × *engelmannii* [Moench]
7 Voss) was more sensitive to the treatments (Kranabetter et al., 2006). The discrepancy might be
8 caused by the lack of representation of relationship between fungi and plants in the model.

9 **4.3 Limitations and challenges**

10 PnET-CN can explicitly simulate the effects of disturbance, pollution, and climate change
11 on forest C dynamics (e.g., Ollinger et al., 2002; Pan et al., 2009; Peters et al., 2013). Despite the
12 capability of the model, the model has some limitations in simulating harvesting effects, and
13 accurate representation of the trajectories of C fluxes and stocks following harvests still remains
14 a challenge.

15 The performance of the model to simulate forest regrowth after harvests is limited by the
16 absence of population and community dynamics associated with regeneration. Most process-
17 based models such as PnET-CN and TEM (Raich et al., 1991) have been developed primarily to
18 simulate C balance for mature forests over the past decades (Landsberg, 2003), resulting in no
19 provision for simulating regeneration such as shrub component and species-specific successional
20 dynamics in these models. Changes in forest composition (e.g., evergreen and deciduous tree
21 species and understory shrubs) along the course of succession are not fully considered by most
22 ecosystem models. PnET-CN does not simulate shrubs and herbs that likely dominate stands in
23 the early successional stage after stand-replacing harvests. The model thus is not able to simulate
24 the particularly high GPP and ER in the young forests where forest canopy has not yet fully
25 recovered.

26 Understory layer is also an important component for mature forest ecosystems in terms of
27 C fluxes and stocks. Misson et al. (2007) reported that understory can contribute 11% (range, 0–
28 39%) of GPP at 10 sites across a wide range of forest type and climate. PnET-CN slightly
29 overestimated overstory LAI for the mature DBF sites and reasonably predicted foliar N
30 concentration compared to satellite-based estimates. The lack of understory layer in the model is

1 possibly responsible for the underestimation of GPP for mature DBF sites. Species competition
2 and cohort methods that have been employed in other models such as ED (Medvigy et al., 2009)
3 and LPG-Guess (Smith et al., 2001) could be used to improve the regeneration and understory
4 components of PnET-CN in the future.

5 Parameter values used in the model were generally derived from specific measurements
6 for a given stand development stage particularly mature forests, although the parameter values
7 likely differ with stand development. For example, the canopy light attenuation constant
8 coefficient is typically measured in mature forests (e.g., Ryu et al., 2008), although the
9 coefficient is known to change with canopy cover (Brantley and Young, 2007). The use of the
10 canopy light attenuation coefficient measured in mature forest for whole forest life simulations
11 could slow down stand development due to the underestimation of photosynthesis in young
12 forests. Understanding the relationship between such parameters and state variables (e.g., LAI) is
13 thus one of the challenges to simulate the effects of stand-replacing harvests on forest C
14 dynamics.

15 Changing climate conditions can also affect the values of some parameters. For example,
16 wood turnover rate (% , tree mortality in terms of biomass losses), to which wood living biomass
17 C and soil organic C are sensitive, could be altered by extreme weather conditions including
18 droughts (Allen et al., 2010; Wang et al., 2012a). Most process-based models are not able to
19 simulate the mechanistic processes associated with tree mortality under changing climate
20 conditions (McDowell, 2011; Wang et al., 2012a), although there is growing interest in the
21 mechanistic modeling of forest mortality (e.g., McDowell et al., 2013; Powell et al., 2013).
22 Recent studies have revealed that climate and disturbance legacies govern forest C dynamics
23 (Magnani et al., 2007; Bond-Lamberty et al., 2013). Future modeling efforts can benefit from
24 improved understanding of the effects of climate change on parameter values that are assumed to
25 be constant in the model.

26 Harvest methods depend on forest types, management needs, and species to be
27 regenerated. For example, selective harvesting or shelterwood system is typically used for
28 hardwoods in Wisconsin (Wisconsin Department of Natural Resources, 2011). Stand-replacing
29 harvesting was assumed for both DBF and ENF chronosequence sites due to the lack of
30 harvesting information and the types of clearing applied to the sites studied. The sensitivity

1 analysis conducted in this study suggests that harvest intensity affects C dynamics in early
2 succession after harvesting. Observations in residuals and post stands after each operation type
3 (e.g., pre-commercial thinning and selective harvesting) are needed to parameterize process-
4 based models for better mechanistic understanding of the harvest effects on forest C dynamics.

5 **5 Conclusions**

6 The PnET-CN model was generally able to simulate the effects of stand-replacing
7 harvests on forest C dynamics (C fluxes and AGC) for two northern temperate forest
8 chronosequences. The predicted dynamics in NEP and AGC following clear-cuts generally
9 follow the hypothesized trajectories, although our simulations show that the decline in NEP was
10 due to relatively stable GPP and gradually increasing ER. Our study also shows that DBF
11 recovered faster (11 years) from net C sources to net sinks and accumulated more C in AGC than
12 ENF. Northern temperate ENF is more vulnerable to stand-replacing harvests than northern
13 temperate DBF. Future research is needed to better understand how respiration components (e.g.,
14 ecosystem and soil respiration) and production components (e.g., overstory and understory)
15 change with forest age and their determinants. Modeling the combined effects of climate change
16 and forest management will benefit from the incorporation of forest population dynamics (e.g.,
17 regeneration and mortality), relationships between age-related model parameters and state
18 variables (e.g., LAI), and silvicultural system into the model. With these improvements, process-
19 based ecosystem models can better simulate regional C balance associated with disturbance
20 regime under changing climate.

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28

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23

1 **Tables**

2 Table 1 Site characteristics for two chronosequences of deciduous broadleaf forests (DBF) and evergreen needleleaf forests (ENF) in
 3 Upper Midwest region of Wisconsin and Michigan, United States.

Site	ID	Location	Plant function type	Dominant species	Year of recent disturbance	AGC (Mg ha ⁻¹ , 2005)	LAI (m ² m ⁻² , 2002)	Data period	Reference
Clearcut young hardwood	YHW	46.72°N 91.25°W	DBF	Aspen, red maple	1999	3.3 (1.3)	0.79 (0.6)	2002	Noormets et al. 2007
Intermediate hardwood	IHW	46.73°N 91.23°W	DBF	Aspen	1984	47.6 (15.6)	3.0	2003	Noormets et al. 2008
Willow creek	WIC	45.80°N 90.08°W	DBF	Sugar maple, basswood, green ash	1875 ^a , 1933	74.9 ^b	5.36 (0.47) ^c	2000-2006	Cook et al., 2008, Curtis et al. 2002
University of Michigan Biological Station	UMBS	45.56°N 84.71°W	DBF	Aspen, white pine, red oak, sugar maple	1920	73.2 (3.1) ^d	3.54 (0.31) ^e	2000-2003	Gough et al. 2008
Young red pine	YRP	46.72°N 91.18°W	ENF	Red pine, jack pine	1993	7.7 (8.3)	0.52 (0.3)	2002	Noormets et al. 2007
Young jack pine	YJP	46.62°N 91.08°W	ENF	Jack pine	1987	4.9 (5.0)	0.93	2004-2005	Noormets et al. 2008
Intermediate red pine	IRP	46.69°N 91.15°W	ENF	Red pine	1980	47.7 (37.3)	3.0	2003	Desai et al. 2008
Mature red pine	MRP	46.74°N 91.17°W	ENF	Red pine, aspen	1939	56.9 (33.1)	2.7 (0.8)	2002-2005	Noormets et al. 2007

4 ^a estimated year of disturbance based on Ameriflux site description in AmeriFlux.
 5 ^b sum of wood and foliage biomass carbon from Curtis et al., 2002.
 6 ^c estimated values based on measurements in 1998 to 2000 and 2002 from Cook et al., 2008.,
 7 ^d value in 2003 from Gough et al. 2008.
 8 ^e calculated based on multi-year (1999-2003) estimations with litter traps from Gough et al. 2008.

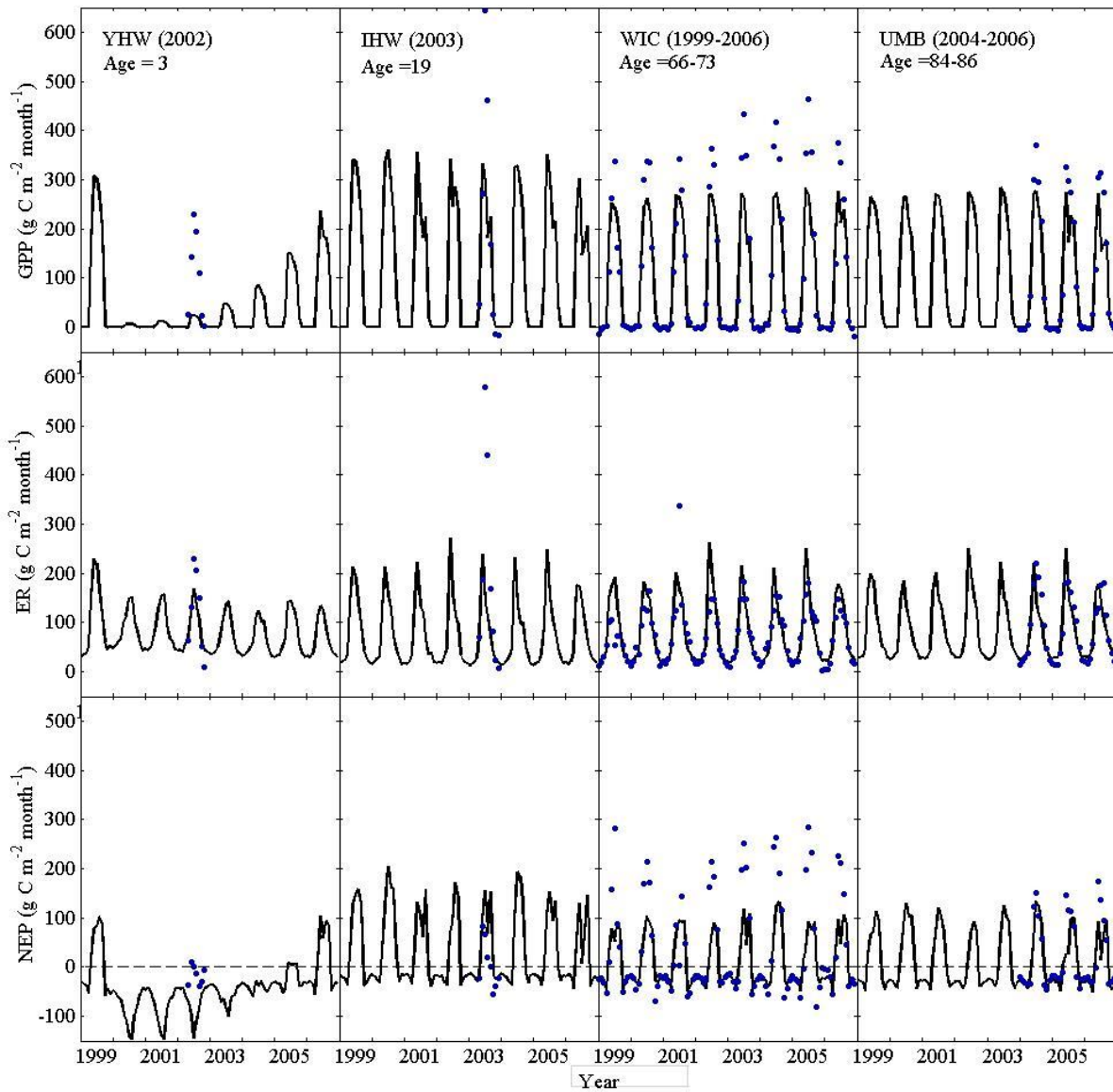
1 Table 2 PnET-CN model performance in monthly carbon fluxes (GPP: gross primary
 2 productivity; ER: ecosystem respiration; NEP: net ecosystem productivity), leaf area index
 3 (LAI), and aboveground carbon stock (AGC) for the two chronosequences.

	NRMSE%^a	<i>d</i>^b	n
<i>DBF</i>			
GPP	10	0.95	147
ER	10	0.92	147
NEP	17	0.81	147
LAI	33	0.97	4
AGC	42	0.95	4
<i>ENF</i>			
GPP	28	0.91	64
ER	37	0.88	64
NEP	46	0.58	64
LAI	29	0.96	4
AGC	37	0.94	4
<i>Overall</i>			
GPP	11	0.94	211
ER	10	0.91	211
NEP	21	0.73	211
LAI	31	0.96	8
AGC	28	0.95	8
Total performance ^c	20	0.90	

4 ^a Normalized root mean square error.

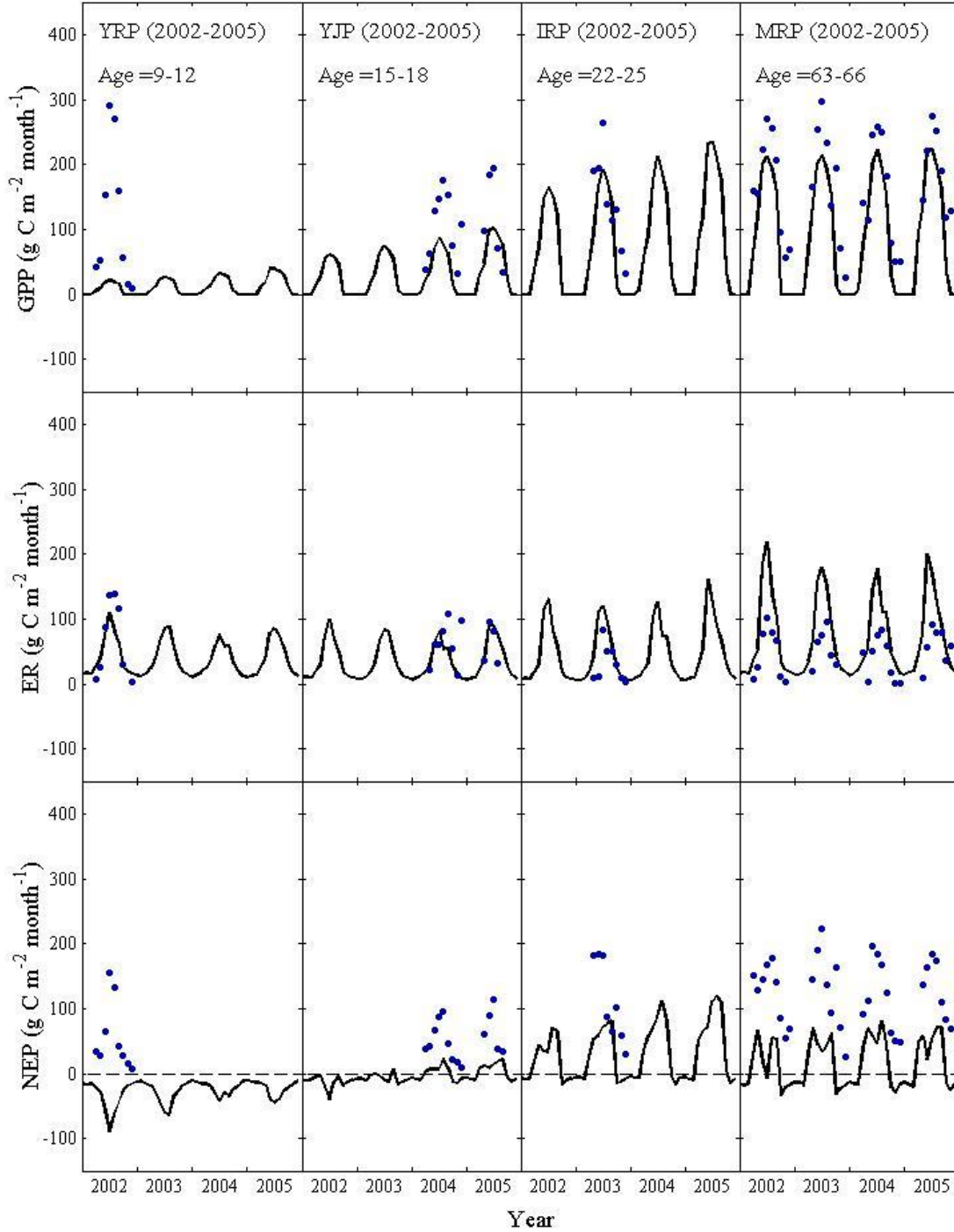
5 ^b Willmott index.

6 ^c Mean value of evaluating statistics for all tested variables.



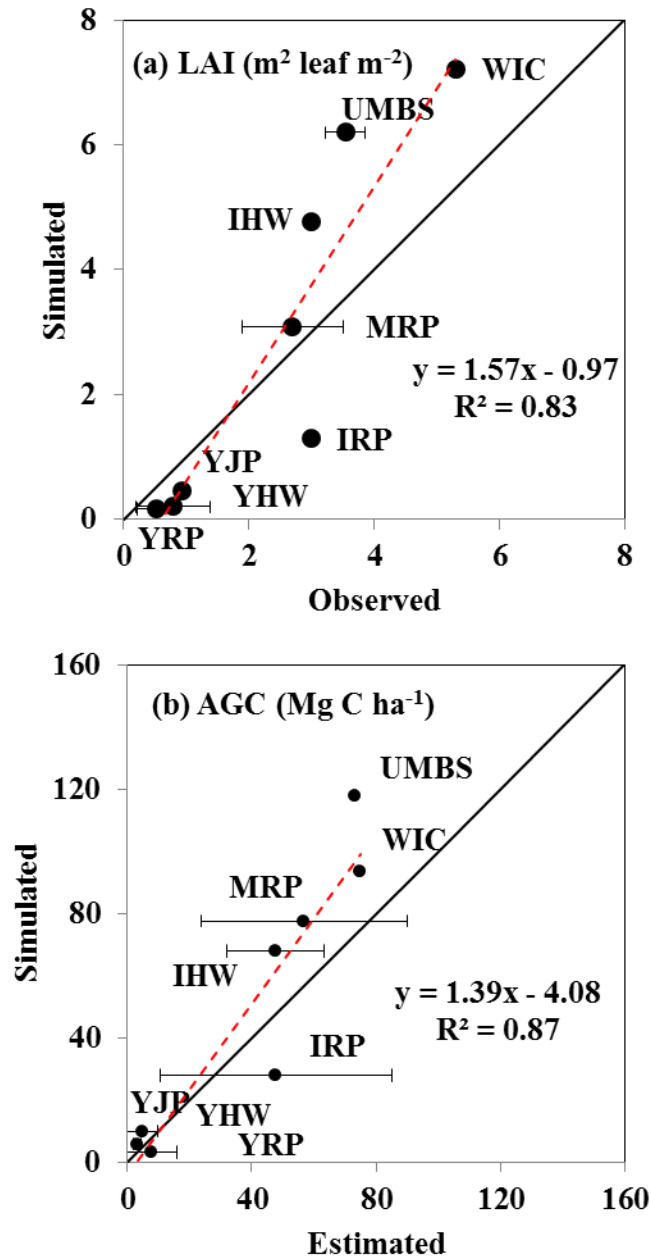
1
 2 Figure 1. Simulated (lines) and observed (symbols) monthly carbon fluxes: GPP, ER, and NEP
 3 for the deciduous broadleaf chronosequence sites from 1999-2007.

4



1
 2 Figure 2. Simulated (lines) and observed (symbols) monthly carbon fluxes: GPP, ER, and NEP
 3 for the evergreen coniferous chronosequence study sites from 2002-2005.

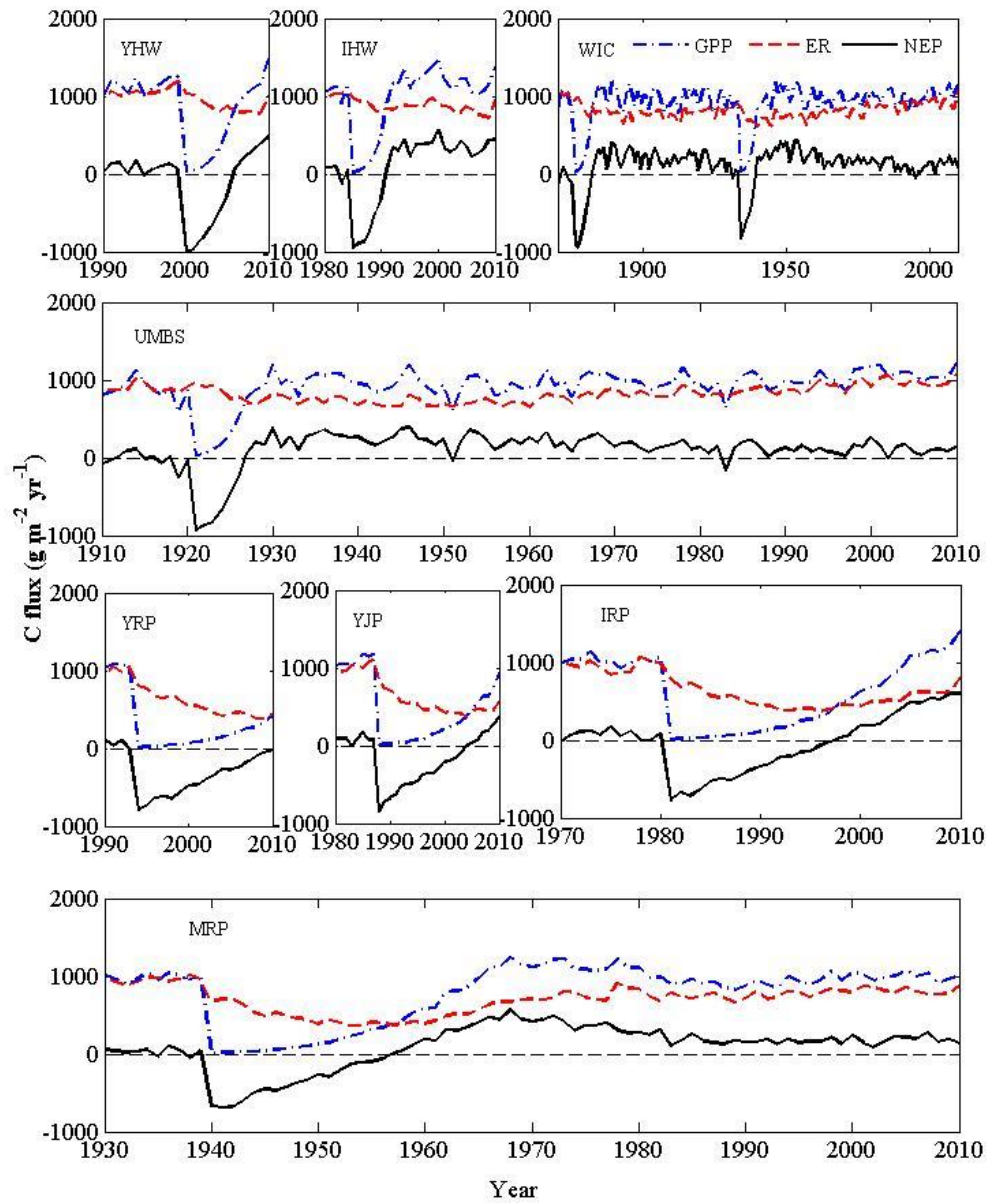
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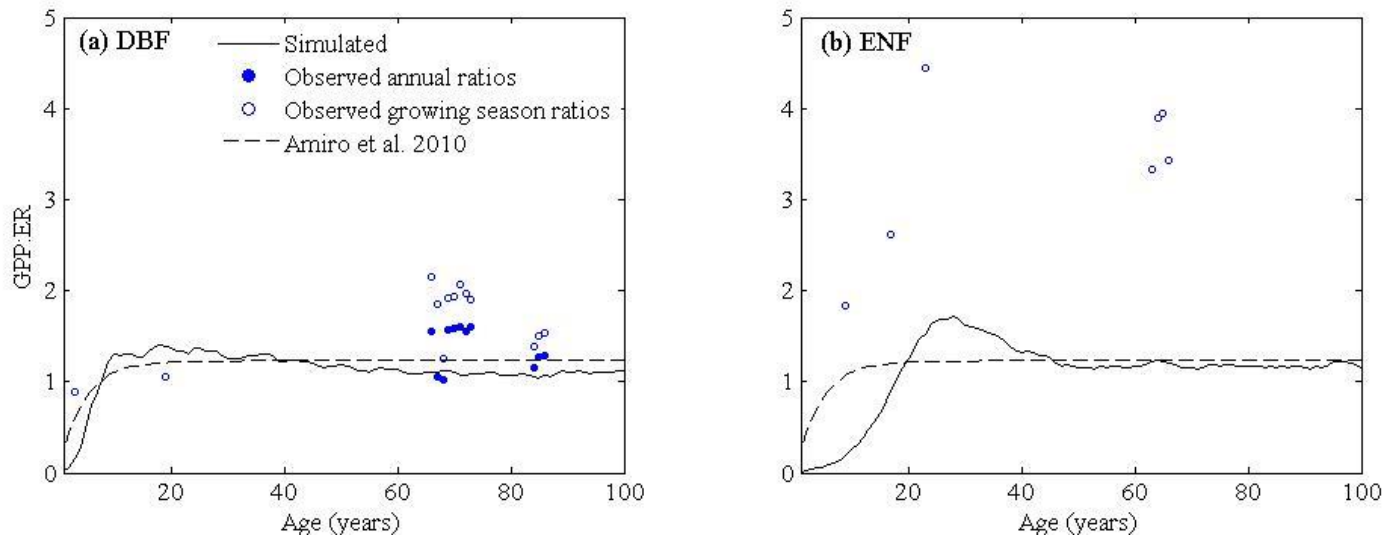
2 Figure 3. Comparisons of simulated and observed (a) leaf area index (LAI) and
 3 aboveground carbon stock (AGC) for all eight sites.

4



1
 2 Figure 4. Simulated trajectories of GPP, ER, and NEP for each site based on the site disturbance
 3 history (Table 1). The time series started from the earliest major disturbance for each site.
 4

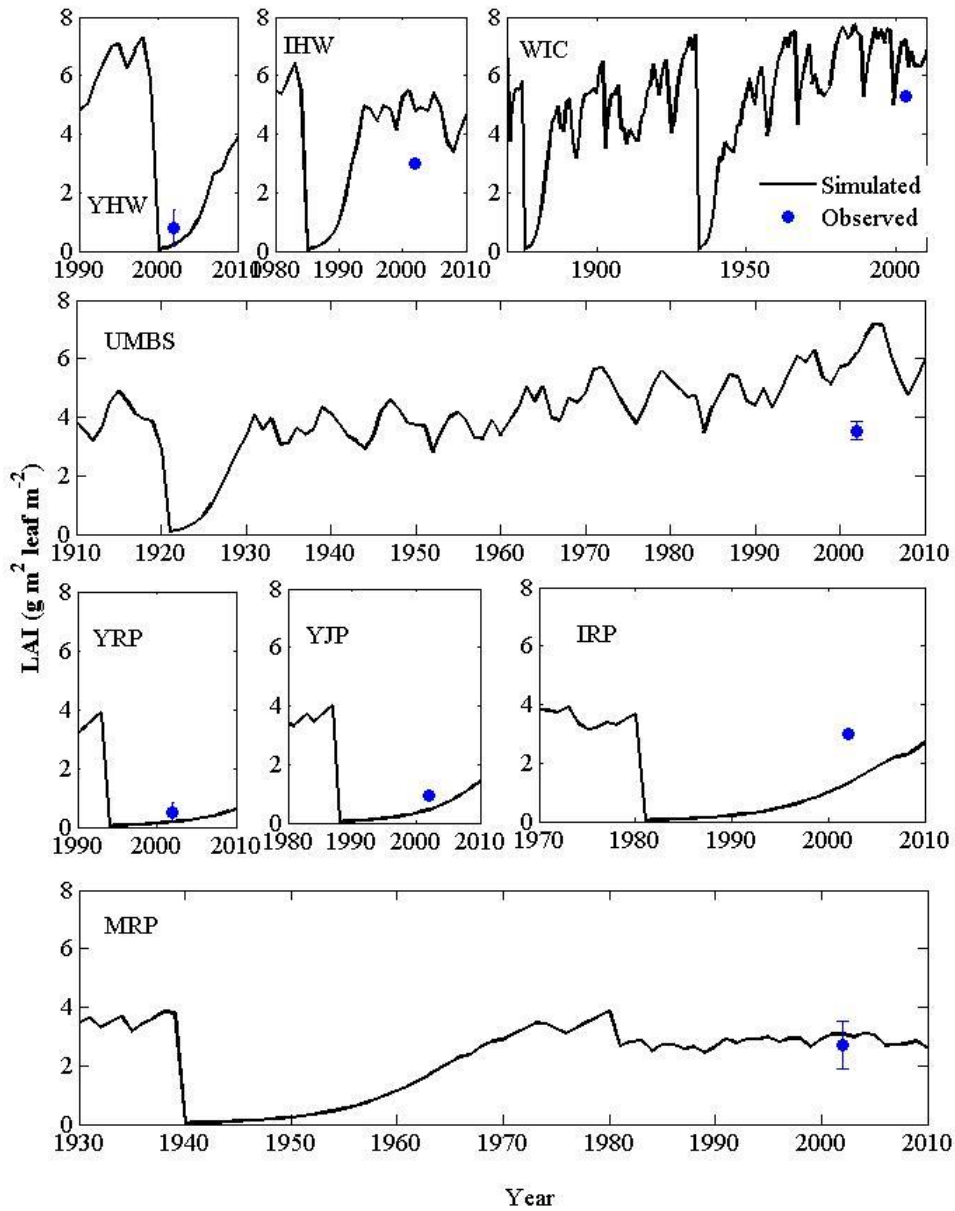
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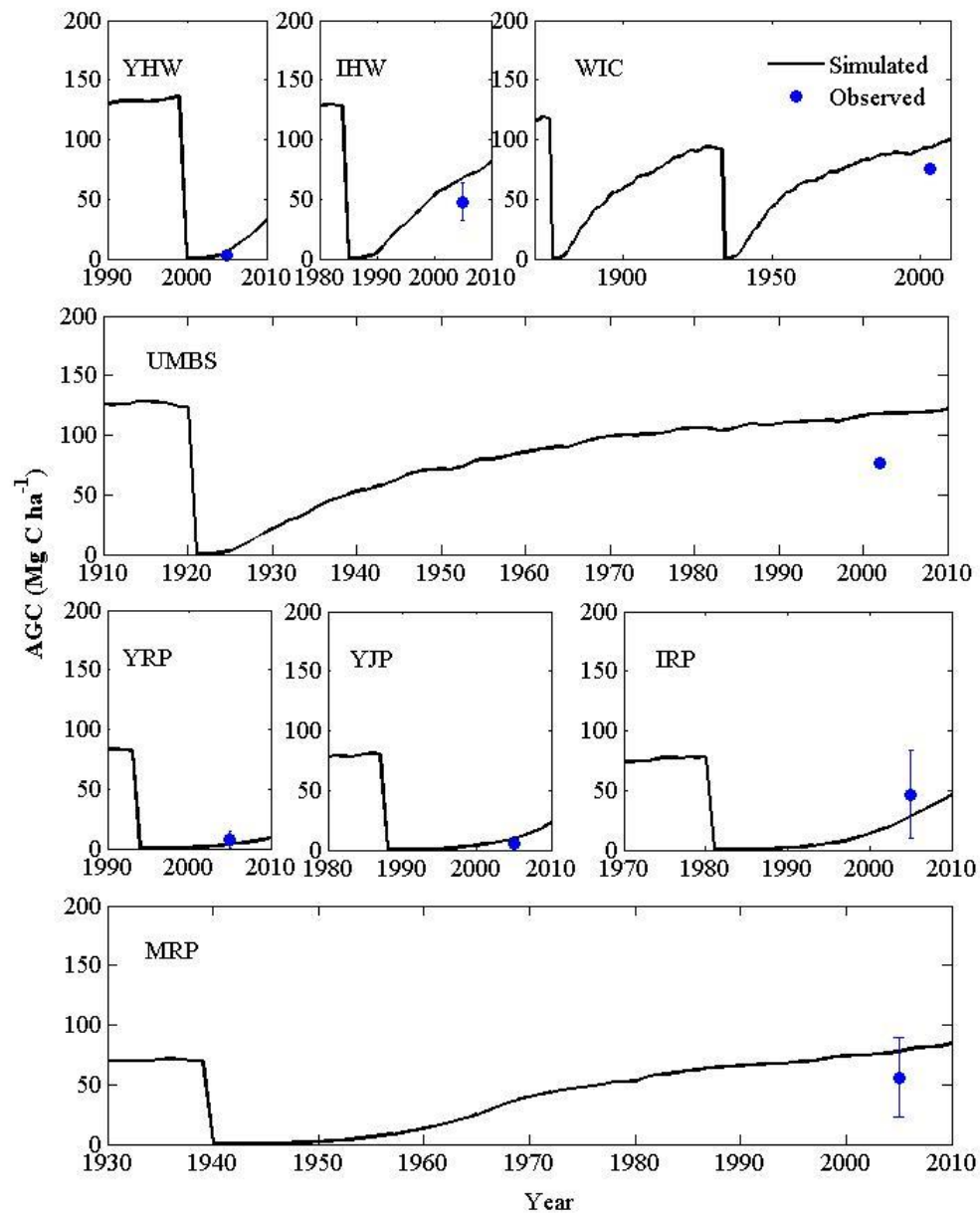
3 Figure 5. Simulated trajectories of the annual GPP/ER ratio with stand age for (a) deciduous
4 broadleaf forests (DBF) and (b) evergreen needleleaf forests (ENF). The dashed line is a fitted
5 curve derived by Amiro et al. (2010) using eddy covariance observations from forest
6 chronosequences in North America. Solid and hollow circles represent measured annual and
7 growing season (May to October) ratios, respectively. The simulated curves were smoothed
8 using a moving average filter with a span of 5.

9



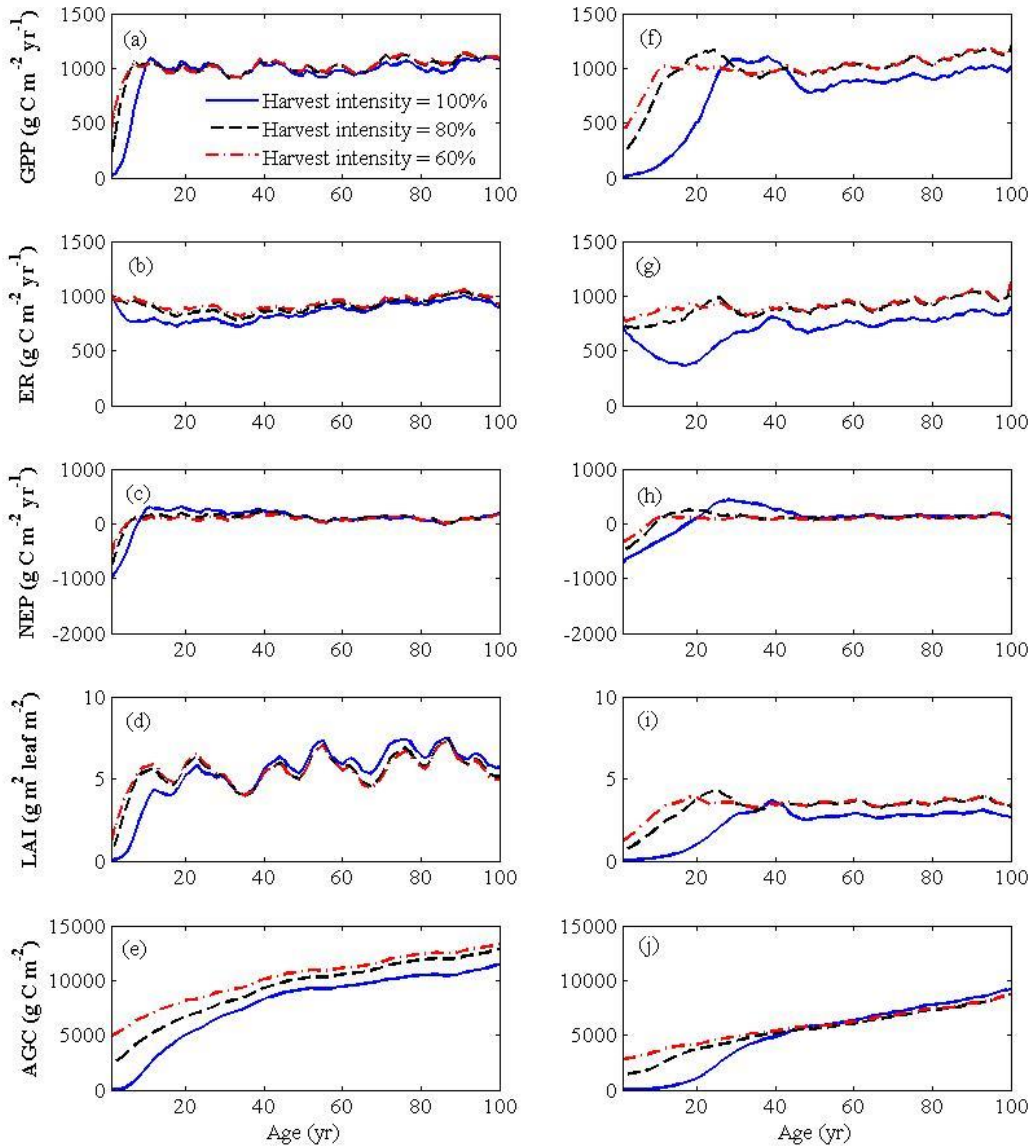
1
 2 Figure 6. Simulated trajectories of LAI for each site based on the site disturbance history (Table
 3 1). The time series started from the earliest major disturbance for each site. Symbols represent
 4 measured LAI.

5



1

2 Figure 7. Simulated trajectories of aboveground biomass carbon (AGC) for each site based on
 3 the site disturbance history (Table 1). The time series started from the earliest major disturbance
 4 for each site. Symbols represent estimated AGC.



1
 2 Figure 8. Sensitivity of carbon fluxes (GPP, gross primary production; ER, ecosystem
 3 respiration; NEP, net ecosystem production) and stand characteristics (LAI: leaf area index;
 4 AGC: aboveground carbon stock) to changes in harvest intensity (reduced by 0.2 and 0.4
 5 compared to 1 for assumed clear-cuts used in the model tests) for (a-e) deciduous broadleaf
 6 forests (DBF) at Willow creek and (f-j) evergreen coniferous forests (ENF) at Mature red pine
 7 site over a 100-yr harvest cycle. The simulated curves were smoothed using a moving average
 8 filter with a span of 5.

9