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2 Carbon consequences of forest disturbance and recovery across the conterminous
3 United States

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21 **Abstract:**

22 Forests of North America are thought to constitute a significant long term sink for atmospheric
23 carbon. The United States Forest Service Forest Inventory and Analysis (FIA) program has
24 developed a large data base of stock changes derived from consecutive estimates of growing stock
25 volume in the US. These data reveal a large and relatively stable increase in forest carbon stocks
26 over the last two decades or more. The mechanisms underlying this national increase in forest
27 stocks may include recovery of forests from past disturbances, net increases in forest area, and
28 growth enhancement driven by climate or fertilization by CO₂ and Nitrogen. Here we estimate
29 the forest recovery component of the observed stock changes using FIA data on the age structure
30 of US forests and carbon stocks as a function of age. The latter are used to parameterize forest
31 disturbance and recovery processes in a carbon cycle model. We then apply resulting
32 disturbance/recovery dynamics to landscapes and regions based on the forest age distributions.
33 The analysis centers on 28 representative climate settings spread about forested regions of the
34 conterminous US. We estimate carbon fluxes for each region and propagate uncertainties in
35 calibration data through to the predicted fluxes. The largest recovery-driven carbon sinks are
36 found in the Southcentral, Pacific Northwest, and Pacific Southwest regions, with spatially
37 averaged net ecosystem productivity (*NEP*) of about 100 g C m⁻² a⁻¹ driven by forest age
38 structure. Carbon sinks from recovery in the Northeast and Northern Lake States remain
39 moderate to large owing to the legacy of historical clearing and relatively low modern disturbance
40 rates from harvest and fire. At the continental scale, we find a conterminous U.S. forest *NEP* of
41 only 0.16 Pg C a⁻¹ from age structure in 2005, or only 0.047 Pg C a⁻¹ of forest stock change after
42 accounting for fire emissions and harvest transfers. Recent estimates of *NEP* derived from
43 inventory stock change, harvest, and fire data show twice the *NEP* sink we derive from forest age

44 distributions. We discuss possible reasons for the discrepancies including modeling errors and
45 the possibility of climate and/or fertilization (CO_2 or N) growth enhancements.

46

47 **1. Introduction**

48 The global imbalance among ocean, industrial, and land use sources/sinks of CO₂ and the
49 amount accumulating in the atmosphere implies significant net CO₂ uptake by the terrestrial
50 biosphere [e.g. *Schimel et al.*, 2001; *Tans et al.*, 1990]. Despite large uncertainty about
51 magnitude and process, analyses tend to point to northern temperate and boreal lands as dominant
52 terrestrial sinks of CO₂ but with considerable controversy regarding attribution to specific regions
53 or continents [e.g. *Bousquet et al.*, 2000; *Fan et al.*, 1998; *Gurney et al.*, 2002; *Kaminski et al.*,
54 2001; *Myneni et al.*, 2001; *Tans et al.*, 1990]. However, some recent work suggests far smaller
55 sinks in northern temperate and boreal lands [*Ito et al.*, 2008; *Stephens et al.*, 2007; *Yang et al.*,
56 2007].

57 Estimates of the conterminous U. S. forest net carbon uptake from the atmosphere range from
58 only 10 to over 200 Tg C a⁻¹ [*EPA*, 2010; *Houghton et al.*, 1999; *King et al.*, 2007; *Pacala et al.*,
59 2001; *Turner et al.*, 1995] in the last 2 decades. Note that here we consider the forest stock
60 change alone rather than the forest sector stock change, where the latter also includes carbon
61 accumulated in wood products (see State of the Carbon Cycle Report [*King et al.*, 2007]).

62 Techniques for estimating forest carbon fluxes at regional to national scales include three
63 approaches. The stock change method is exemplified in the US report to the United Nations
64 Framework Convention for Climate Change [e.g. *EPA*, 2008] which uses US Forest Service
65 Forest Inventory and Analysis (FIA) data on sequential measurement of tree diameters and/or
66 wood volumes for about 100,000 forest plots at 5-20 year intervals. Allometric and biomass
67 expansion factors are used to convert volume into forest carbon stocks. The rate of carbon uptake

68 is then estimated as the difference between sequential measurements divided by the number of
69 years in the interval.

70 Another technique for estimating forest carbon sinks combines estimates of the stand age
71 structure of forests with age-specific carbon accumulation rates, termed the “age-accumulation”
72 approach in this work. These carbon accumulation rates are inferred from carbon stocks as a
73 function of age [e.g. *Houghton, 1999*], known as yield tables in forestry literature, and may be
74 derived empirically from inventory estimates of stand volume and age or from a process oriented
75 dynamic growth model. Finally, forest carbon sinks have been estimated from process models
76 that account for the effects of climate variability and CO₂ and nitrogen fertilization but not
77 necessarily for land use and disturbance processes [e.g. *Schimel et al., 2000*]. These effects are
78 fully contained in the stock change method because it relies on contemporary changes in stocks,
79 but the age-accumulation approach relies on a historical characterization of carbon stock
80 accumulation and thus misses some of the contemporary influences (see Auxiliary 4).

81 Forest stock changes result from the sum of net ecosystem productivity (*NEP*), fire losses, and
82 harvest (see Figure 1). Significant decreases in harvest and fire have not been observed over the
83 past few decades so speculation as to the mechanisms underlying the stock increases have focused
84 more on growth enhancement from either climate change or fertilization with elevated carbon
85 dioxide or nitrogen [*Houghton, 1999; McGuire et al., 2001; Nemani et al., 2002; Pan et al., 2009;*
86 *Schimel et al., 2000; Zhou et al., 2003*] and on forest growth from post-disturbance recovery or
87 fire suppression [*Caspersen et al., 2000; Hurtt et al., 2002; Pacala et al., 2001*]. Though the
88 growth enhancement hypothesis has been challenged by *Caspersen et al.* [2000] using forest
89 inventory data, others have argued that plausible rates of growth enhancement cannot be detected
90 using existing inventories [*Joos et al., 2002*] and recent work presents observational evidence

91 supporting a large climate change or fertilization induced sink [Cole *et al.*, 2010; McMahon *et al.*,
92 2010; Thomas *et al.*, 2009].

93 Disturbed forests, if not converted to another land cover type, have the potential to regrow,
94 recover, or even surpass pre-disturbance carbon stocks over decades to several hundred years.
95 The long-standing dogma of the carbon source/sink dynamics for stand-replacing disturbance
96 involves a rapid pulse emission followed by sizeable net uptake that gradually declines [Koerner,
97 2003; Odum, 1969]. This pattern is broadly supported by chronosequence observations of carbon
98 stocks [Bond-Lamberty *et al.*, 2004; Gough *et al.*, 2007; Pregitzer and Euskirchen, 2004; Richter
99 *et al.*, 1999; Thornton *et al.*, 2002] and forest-atmosphere net CO₂ exchange [Amiro *et al.*, 2011;
100 Barford *et al.*, 2001; Goulden *et al.*, 2011; Law *et al.*, 2003; Schwalm *et al.*, 2007], but the precise
101 post-disturbance carbon dynamics vary by forest type and climate and this detail remains poorly
102 characterized.

103 The analysis reported here attempts comprehensive assessment of the carbon consequences of
104 past and present forest disturbance and recovery across the conterminous United States. We ask
105 if the forest age structure of the conterminous US forests accounts for the stock changes reported
106 by the FIA. Our approach utilizes the national forest inventory data (and uncertainties) to
107 constrain the forest disturbance and recovery processes represented in an ecosystem carbon cycle
108 model to obtain regional and national estimates of carbon consequences. The basic method can
109 be described as having two main steps. First, we derive forest type and climate specific post-
110 disturbance *NEP* trajectories by fitting a first-order terrestrial carbon cycle model (CASA, [Potter
111 *et al.*, 1993; Randerson *et al.*, 1996]) to grow wood stocks consistent with FIA data. Second,
112 these characteristic trajectories are applied to landscapes with forest age maps obtained from FIA
113 age distributions to derive maps of *NEP* and biomass. As such, our approach corresponds to the

114 age-accumulation method for estimating forest carbon sinks as described above. Results
115 represent carbon dynamics of forested ecoregions across the conterminous US to provide a
116 continental-scale view of forest recovery from past disturbances. In addition, we formally
117 propagate the uncertainty in FIA age-biomass trends using a Monte Carlo approach, as well as
118 examine to what degree results are sensitive to uncertainty in the model's parameterization of
119 carbon turnover time, and dependence on light, moisture, and temperature. Discrepancies
120 between FIA estimates of stock changes and those from our age-accumulation modeling are
121 assessed in terms of modeling errors and potential growth enhancements above and beyond
122 recovery, similar to *Houghton* [2003].
123

124 **2. Methods**

125 **2.1 Overview**

126 The core of our approach is to estimate the frequency (F) of land area in a region (A_{reg}), as
127 well as the flux or stock of carbon (Q) each within strata of stand age, forest type (e.g. Aspen-
128 Birch), and site productivity (high or low) (denoted with a, f, p subscripts). Regions are defined
129 according to the Resource Planning Act Assessment by the US Forest Service. From this we
130 calculate the regional mass flux or stock ($Q_{reg,s}$) for a particular climate setting (subscript s) within
131 each region, as well as its uncertainty (δ , described further below), according to
132

$$133 \quad Q_{reg,s} = \sum_a \sum_f \sum_p Q_{afp} F_{afp} A_{reg}$$

134 (1),

135

136 where F is the frequency of forest area adjusted to sum to unity over the three strata and obtained
 137 from the regional FIA samples of the area of forest land as described in Section 2.2, A_{reg} is the
 138 total forested area in the region, and subscripts are: a for stand age, f for forest type group, and p
 139 for productivity class. The work reported here is part of a larger project to incorporate stand age
 140 derived from Landsat time series data. In this parallel effort, specific scenes for Landsat time
 141 series were obtained from a statistically rigorous sampling procedure of forest type spatially
 142 dispersed within Eastern and Western regions [Goward *et al.*, 2008]. Here we use the climate
 143 (temperature, precipitation, incident solar radiation) and phenology for each scene (Figure 2) to
 144 simulate fluxes and stocks for each forest type and productivity class within the scene. The scenes
 145 within a region are generally good representations of the region except for the Pacific Southwest
 146 where coastal forests are not well represented. The scene level fluxes are then aggregated to
 147 regional forest fluxes and stocks by averaging across the number of climate settings (scenes, N_s)
 148 in a region as

149

$$\begin{aligned}
 Q_{reg} &= \frac{1}{N_s} \sum_s Q_{reg,s} \\
 \delta Q_{reg} &= \frac{1}{N_s} \sum_s \delta Q_{reg,s}
 \end{aligned}
 \tag{2},$$

151

152 and conterminous US estimates (subscript nat) are obtained from the sum over regions

153

$$\begin{aligned}
 Q_{nat} &= \sum_{reg} Q_{reg} \\
 \delta Q_{nat} &= \sum_{reg} \delta Q_{reg}
 \end{aligned}
 \tag{3}.$$

155

156 We note that our estimates do not account for possible changes in forest carbon due to changes in
157 forest area, though in the Discussion section we explain why this is unlikely to contribute a large
158 carbon source or sink given the rates of current-day net land conversion.

159 The relationship between fluxes and stocks can be diagramed as shown in Figure 1. The so-
160 called forest sector sources/sinks refer to the net flux between the atmosphere and forest stocks
161 plus wood products stocks. The inventory approach to calculating the net forest-atmosphere flux
162 involves a measured change in carbon stocks over a specified period. A change in forest carbon
163 stocks can occur because of changes in the physiological fluxes of photosynthesis and ecosystem
164 respiration (balanced as *NEP*), as well as changes in disturbance for example by fire or harvest.
165 *NEP* can then be inferred as the difference between ΔC_{stocks} and removals from fire and harvest.
166 The net forest sector flux to the atmosphere is the sum of ΔC_{stocks} and $\Delta C_{wood\ products}$. This
167 approach, used in national reports to United Nations Framework Convention on Climate Change,
168 derives $\Delta C_{wood\ products}$ from independent harvest records and empirical decay constants for wood
169 products and landfills.

170 Our approach is to calibrate our modeled biomass as a function of age using forest inventory
171 data. We then apply the biomass and associated *NEP* from forest disturbance and recovery to the
172 landscape based on the forest area reported by the FIA within strata of age, forest types and
173 productivity classes within each region. In our modeling framework an important driver of
174 ΔC_{stocks} is net primary production (*NPP*), and the turnover times of wood and detrital pools. *NPP*
175 allocated to leaves and fine roots is quickly decomposed and cannot represent a persistent (>
176 decadal) sink. The turnover rates of wood and its immediate detrital pool, coarse woody debris,
177 are much slower, on the order of decades, and thus able to account for long-term net carbon fluxes
178 (on the order of a century). Fluxes from large stocks of slowly overturning soil pools are also

179 slow to respond to disturbance. By the time these large soil pools are affected by disturbance,
180 recovery may have already occurred. This phenomenon is expressed as a low sensitivity of *NEP*
181 to the slow turnover pools in recovering forests (see Auxiliary Material, Auxiliary 1). Of course
182 the slow soil pools are a significant source or sink in conditions where changes in fluxes into the
183 slow pools are large and longer term such as in permanent conversion from or to forest. This
184 approach allows us to map *NEP* from recovery, one of the key atmospheric flux components
185 needed to understand source/sink processes. *NEP* is a purely biological flux dependent on
186 photosynthesis and respiration alone. Fluxes out of the forest arising from harvest or fire combine
187 with *NEP* to produce net biome productivity (*NBP*) which is equivalent to ΔC_{stocks} . Note that we
188 have neglected the generally smaller fluxes that contribute to *NBP* such as lateral fluxes of
189 carbonate and organic matter in liquid form as well as volatile organic carbon emissions (see
190 *Chapin III et al.* [2006]).

191

192 **2.2 Data Sources and Modeling**

193 Flux trajectories are derived by fitting forest growth, mortality and shedding, and allocation
194 parameters within the Carnegie-Ames-Stanford Approach (CASA) carbon-cycle process model
195 [*Potter et al.*, 1993; *Randerson et al.*, 1996] to accumulate carbon in aboveground wood biomass
196 consistent with forest inventory data. Productivity in CASA is represented with a light use
197 efficiency approach in which *NPP* is proportional to the fractional absorption of
198 photosynthetically active radiation (f_{PAR}) times an efficiency term modulated by environmental
199 conditions. *NPP* is allocated to leaves, roots, and wood which have specific turnover rates that
200 reflect the delivery of carbon to nine detrital pools on the surface and in the soil. These pools
201 decompose at specific turnover rates that are also modulated by environmental conditions.

202 Disturbance causes *NPP* to initially decrease, and removes or transfers carbon between live and
203 detrital pools, the atmosphere, and forest harvest. In this implementation, we adjust the default
204 rate of productivity to match carbon accumulation observed in age-accumulation trajectories from
205 forest inventory data.

206 Inventory data were obtained from the FIA field plots (FIA Database Version 4), providing
207 means and sampling errors for two attributes: 1) all live, oven-dry aboveground wood biomass,
208 and 2) area of forest land. The quotient of these attributes provides biomass per unit area. Each
209 attribute was sampled within strata of forest type group (28 classes), age (20 year age classes to
210 200+ years), and lumped into high and low productivity classes, defined as 120 to >225 cubic feet
211 $\text{acre}^{-1} \text{ annum}^{-1}$ and 20 to <120 cubic feet $\text{acre}^{-1} \text{ annum}^{-1}$ respectively. Inventory samples were
212 drawn for regions defined by the Resource Planning Act Assessment by the US Forest Service
213 that divides the conterminous U.S. into the Northeast (NE), Southeast (SE), Northern Lakes States
214 (NLS), South Central (SC), Northern Prairie States (NPS), Rocky Mountain North (RMN), Rocky
215 Mountain South (RMS), Pacific Southwest (PSW), and Pacific Northwest (PNW) region (Figure
216 2). FIA data on forest carbon and area that are available via World Wide Web download include
217 variances for each. However these variances cannot be exactly combined to estimate uncertainty
218 because of unknown covariance between carbon stock and area [*Bechtold and Patterson, 2005*].
219 Statisticians from the FIA (Charles Scott and colleagues, USFS National Inventory and
220 Monitoring Applications Center) processed the national plot data to provide our study with
221 custom products that we employed in this analysis, namely the aboveground live wood biomass
222 per unit area and its variance for each major forest type, age cohort, productivity class, for each
223 region shown in Figure 2. We confirmed that the data in this custom delivery were nearly

224 identical to those obtained from other web-based data servers maintained and made available by
225 the FIA.

226 For this implementation we drive the CASA model with the f_{PAR} from a smoothed version of
227 the MODIS MOD15A2 product [Nightingale *et al.*, 2009] for each forest type group as well as
228 climatological seasonality of monthly weather using NASA Goddard Institute of Space Sciences
229 (GISS) air temperature anomalies [Hansen *et al.*, 1999] added to a temperature climatology
230 [Leemans and Cramer, 1991], GISS solar radiation [Zhang *et al.*, 2004], and Global Precipitation
231 Climatology Project (GPCP) precipitation [Adler *et al.*, 2003]. These meteorological driver data
232 were sampled at the 1-degree scale while f_{PAR} was provided at 1 km resolution then averaged for
233 each forest type within each of the 28 simulation climate domains. As such, we obtain carbon
234 flux trajectories for each combination of simulation domains ($n = 28$), forest-type group ($n = 3$ to
235 10), and productivity class ($n = 2$). Forest type group is specified at a 0.01 degree resolution
236 obtained from Zhu and Evans [1994] (<http://www.fia.fs.fed.us/library/maps/>). Grid cell-level
237 fractions of forest land in high and low productivity classes for each forest type and stand age
238 within each region are specified from county level FIA data.

239 We modified CASA to capture disturbance impacts on the carbon cycle as follows. The post-
240 disturbance decline and ensuing recovery of NPP and fractional allocation to wood (τ) are
241 modeled as:

242

$$243 \quad NPP(t) = NPP_{max}(1 - ce^{-kt}) \quad (4),$$

244

$$245 \quad \tau = \min[1, (t - 1) / 8 \text{ years}] / 3 \quad (5),$$

246

247 where t is years since disturbance, NPP_{max} is the climatologically averaged net primary
248 productivity independent of a disturbance legacy, c ($=1.5$) determines the magnitude of
249 disturbance-induced reduction in NPP , k ($=0.8$) determines the rate of NPP recovery, and min is
250 the minimum operator. We introduced this dynamic recovery of NPP after disturbance based on
251 the well documented recovery of NPP [e.g. *Amiro et al.*, 2000; *Hicke et al.*, 2003]. The dynamics
252 of allocation were intended to capture initial investment of NPP into herbaceous biomass with
253 increasing allocation to woody vegetation with age [e.g. *Jokela et al.*, 2004; *Law et al.*, 2002].

254 In order to parameterize the amount of biomass killed by a disturbance we adopt the following
255 treatment. Regardless of the pre-disturbance biomass, we set the post-disturbance biomass to
256 50% of the aboveground live wood biomass reported in the 0-20 year age class. This constrains
257 early regrowth to pass through the youngest age-class in the FIA sample. We then estimate the
258 corresponding fraction of live wood, leaves, and roots killed based on the ratio of their abundance
259 prior to disturbance relative to those immediately after disturbance. Eighty percent of the
260 disturbance-killed aboveground wood and all of the disturbance-killed leaves are assumed to be
261 taken off site and entrained into wood products or promptly combusted and are collectively
262 accounted for as “removals” (fire and harvest), akin to the treatment by *Turner et al.* [1995]. The
263 remaining 20% of disturbance-killed aboveground wood is subject to on-site post-disturbance
264 decomposition as it enters the coarse woody debris pool, also consistent with *Turner et al.* [1995].
265 Disturbance-killed roots decompose on-site, for which 30% of dead coarse roots are assumed to
266 enter a belowground coarse woody debris pool, and 70% of dead coarse roots and all dead fine
267 roots enter the soil metabolic and structural pools, broadly consistent with results presented in
268 [*Gough et al.*, 2007; *Meigs et al.*, 2007]. We note that these and other prescriptions are uncertain,
269 likely vary among disturbance and forest types, and are the subject of ongoing research. In

270 summary, biomass killed in a disturbance event is the difference between pre-disturbance biomass
271 and 50% of the 0-20 year biomass reported by the FIA data. Of the killed biomass, 80% of
272 aboveground wood and all leaves are removed (via harvest or fire) and 20% of the killed
273 aboveground wood enters the coarse woody debris pool. The belowground wood and roots killed
274 by disturbance remain on site to decompose. Figure 3 offers an example, in which aboveground
275 biomass is reduced to $2.5/30 \text{ kg C m}^{-2}$, or <10%, and 80% of this 90% reduction in biomass is
276 assumed to be removed (harvest or fire) while the other 20% is left to decompose on site.

277 With this approach it is then possible to estimate biomass removals as:

$$278 \quad R = A_1 B_{pre} (1 - f_{left}) \quad (6),$$

279 where A_1 is the area of forested land assigned a stand age of one year based on the FIA age
280 histogram, B_{pre} is the pre-disturbance aboveground biomass, and f_{left} (=0.8) is the fraction of
281 biomass left to decompose on-site. Each of these varies by forest type, region, and productivity
282 class. This estimate is subject to errors in the area of forest assigned to this young age class, the
283 age of forests prior to disturbance and correspondingly the biomass pre-disturbance, and
284 uncertainty in the fraction of biomass in disturbed forests that is taken off-site as wood products.
285 Removals from non-stand replacing harvests are not considered in this approach but later in the
286 Discussion section we attempt to quantify the impacts of this assumption.

287 The next step in our model parameterization involves calculating the wood production – wood
288 age pair that allows the best match to the inventory data of aboveground stock recovery, with the
289 following multi-step procedure. First, we calculate a target aboveground live wood biomass (B^* ,
290 in g C m^{-2}) from the mean in the 100 to 200 year old age classes, including successively younger
291 age classes in 20 year increments to ensure a minimum of two samples. The target age (A^* , in
292 years) is obtained from the average of old classes sampled to derive B^* . Second, we approximate

293 the rate of annual aboveground live wood biomass production (P_w , in $\text{g C m}^{-2} \text{ a}^{-1}$), which is a
 294 function of NPP and wood allocation, that would be required to obtain B^* by A^* for a range of
 295 possible wood turnover times (A_w) spanning 30 to 300 years in increments of 10 years by solving
 296 a simplified integral form of the differential equation for biomass with time ($dB/dt = P_w - B/A_w$) to
 297 yield:

298

$$299 \quad P_w = \frac{B^*}{A_w(1 - e^{-\frac{A^*}{A_w}})} \quad (7).$$

300

301 Thus, we obtain an array of possible P_w - A_w pairs that would grow the target biomass by the target
 302 age. In a few particular cases this approach yielded implausible wood ages, but with negligible
 303 consequence for the scales of analyses presented in this study. The third step is to select the pair
 304 that provides a biomass recovery curve most like the inventory sample assessed as that which
 305 minimizes the sum of squared error between modeled and sampled aboveground live wood
 306 biomass. Modeled biomass is calculated at the sample ages (t , in years) according to:

307

$$308 \quad B(t, A_w) = B_0 e^{-\frac{t}{A_w}} + P_w A_w (1 - e^{-\frac{t}{A_w}}) \quad (8),$$

309

310 where B_0 is an assumed initial biomass of 200 g C m^{-2} . Lastly, we linearly rescale the model's
 311 default monthly NPP values to provide an annual total NPP_{max} inferred from the fitted rate of P_w ,
 312 as:

$$313 \quad NPP_{max} = \frac{P_w}{\tau\alpha} \quad (9),$$

314 where τ ($=1/3$) is the allocation of NPP to wood and α ($=0.75$) is the fraction of this that is
315 allocated to the aboveground wood pool (stems and branches) instead of belowground (coarse
316 roots).

317 Following determination of P_w and A_w parameters, characteristic carbon flux trajectories
318 (Q_{afp}) are developed from, first, a 1000 year spin-up to steady-state carbon pools. This is followed
319 by a disturbance prior to the disturbance of interest with 75 years of regrowth for all forest types
320 except loblolly pine and longleaf / slash pine (30 years) and Douglas-fir (200 years). The age of
321 trees at harvest is set to be just older than the typical peak in age histograms reported by the FIA
322 (see Auxiliary 2, Figure A2.3), except where harvest rotations are known to be short (SE and SC
323 pines), or where harvest over previous decades tended to target old growth forests with high
324 economic value (Douglas-fir [*Cohen et al.*, 2002]). This ‘pre-disturbance’ is important in that it
325 establishes the amount of live carbon subject to disturbance-induced disposition, meaning taken
326 off-site as removals or decomposing on-site. Finally, we simulate the most recent disturbance
327 after which we allow 200 years of regrowth to characterize carbon dynamics with stand
328 development. These procedures result in a group of carbon stock age trajectories analogous to
329 yield tables.

330 We have not modified CASA’s default treatment of heterotrophic respiration emerging from
331 microbial decomposition of soil and litter carbon and associated transfers among carbon pools.
332 The general equation for the rate of heterotrophic respiration from a specific carbon pool is:

333

$$334 \quad Rh_{pool} = C_{pool} k_{pool} W_{resp} T_{resp} M, \quad (10),$$

335

336 where C_{pool} is the amount of carbon in a pool, k_{pool} is the pool-specific decay rate constant, W_{resp}
337 and T_{resp} control how respiration depends on soil moisture and temperature states, and M is the
338 carbon assimilation efficiency of the microbes. Total heterotrophic respiration is the sum of that
339 from each of the nine detrital pools.

340

341 **2.3 Uncertainty Analysis**

342 A formal propagation of uncertainty from sampling errors (coefficient of variation, CV) for
343 forested area (± 10 to 100%) and total aboveground live biomass (± 10 to 100%), and volume to
344 carbon conversion ($\pm 7\%$) are all included. The uncertainty in inventory aboveground live
345 biomass per unit area is propagated to the predicted fluxes and aboveground live biomass with a
346 Monte Carlo procedure analogous to Tier 2 uncertainty estimation in the IPCC Good Practice
347 Guide [IPCC, 2000]. The model was fit to 25 different biomass regrowth trajectories, where each
348 trajectory was generated from random samples of the normally distributed aboveground live
349 wood biomass for each age class (25 draws of biomass per unit area from each of 10, 20-year age
350 classes). Forcing the fitted trajectory to conform to the assumption that biomass increases
351 monotonically and saturates with age strongly constrains the resultant age-accumulation curves
352 and their variances (Figure 3). An additional 7% uncertainty is used to account for tree volume to
353 carbon conversion [Smith and Heath, 2001]. Put together this method involved over 130,000
354 simulations of age-dependent dynamics of forest carbon fluxes and stocks. The uncertainty of
355 forest area and aboveground live biomass per unit area is obtained from the FIA data.

356 As shown in Equations 1-3 above, independent uncertainties in the product of flux or stock

357 with area are combined as $\delta Q_{total} = \left(\frac{\delta Q_{afp}^2}{Q_{afp}^2} + \frac{\delta A_{afp}^2}{A_{afp}^2} \right)^{1/2}$ [Taylor, 1997]. We adopt a conservative

358 assumption of non-random error propagation for which uncertainty is additive over forest types,

359 productivity classes, and ages, and also additive spatially for a simulation domain, a region, or the
360 nation. This uncertainty aggregation is analogous to a Tier 1 uncertainty described in the IPCC
361 Good Practice Guidance [IPCC, 2000].

362 Uncertainty in *NEP* also derives from model structure (not analyzed) as well as model
363 parameterization of light, moisture, and temperature sensitivity of heterotrophic respiration and/or
364 *NPP* expressed in the CASA model. As described in Auxiliary Material, Auxiliary 1 Section 1,
365 we analyzed *NEP* responses to a 2% increase of six representative parameters including the
366 maximum light use efficiency, moisture dependence of *NPP*, optimal temperature for *NPP*,
367 turnover time of the slow soil carbon pool, and both the Q10 and moisture dependence of
368 heterotrophic decomposition of soil carbon. We use a 2% change in parameter value in order to
369 obtain a detectable response in *NEP* but for ease of discussion the sensitivities are divided by two
370 and expressed as % change in *NEP* for a 1% change in parameter value (see Auxiliary Material,
371 Auxiliary 1).

372

373 **3. Results**

374 **3.1 Carbon Trajectories**

375 Using CASA as a controlled growth model accurately reproduces the accumulation of
376 aboveground forest carbon stocks with time since a stand replacing disturbance as informed by
377 FIA data (Figure 3), imposing a powerful, albeit partial, observational constraint on net ecosystem
378 carbon flux trajectories with stand age. Additional data on litter, woody debris and soil carbon
379 dynamics would provide much needed additional constraints on estimated ecosystem C dynamics.
380 More rapid regrowth of aboveground stocks in the high productivity class causes higher
381 amplitude trajectories for carbon stocks and fluxes (Figure A2.1, A2.2, Auxiliary 2) with larger

382 post-disturbance sources that give way to stronger sinks with ensuing forest regrowth. The Monte
383 Carlo simulation approach provides an envelope of trajectories (Figure 3) that enables formal
384 uncertainty propagation through all scales of the analysis (regional forest types to conterminous
385 U.S. forestlands). Absolute uncertainty surrounding *NEP* tends to peak where forest uptake is
386 maximum (peak *NEP*) and then diminishes with forest age (Figure 3). An important exception,
387 not shown in Figure 3, is the often large uncertainty in carbon emission in the years immediately
388 following disturbance; large because of variation in the pre-disturbance carbon stocks and the
389 amount of dead wood that decomposes on-site. The timing of *NEP* crossover from source to sink
390 is surprisingly insensitive to variability in biomass accumulation (not shown), and generally
391 occurs at ages <20 years (e.g. Figure 3 and Figure A2.1, A2.2, Auxiliary 2) consistent with many
392 reported chronosequence fluxes [e.g. *Bond-Lamberty et al.*, 2004; *Gough et al.*, 2007; *Goulden et*
393 *al.*, 2011; *Law et al.*, 2004; *Litvak et al.*, 2003; *Noormets et al.*, 2007; *Pregitzer and Euskirchen,*
394 2004]. Patterns of post-disturbance uptake of carbon in regrowing forests vary widely across
395 regions of the conterminous U.S. as well as by forest type group and productivity class (Figure
396 A2.1, A2.2, Auxiliary 2). Forest inventory data describing the recovery of aboveground live wood
397 biomass carbon with stand development act as a strong constraint on the modeled carbon cycle
398 including the rates of litter and soil carbon turnover and decay.

399 Our analysis of the sensitivity of the model to parameters revealed that nearly all of the
400 sensitivities are less than 1% indicating general dampening of parameter perturbations and
401 suggesting that uncertainties in these parameterizations do not expand as they propagate through
402 to modeled *NEP* (see Auxiliary 1, Table A1.1). Model structure and parameter uncertainties are
403 not included in our analysis but are expected to add about $\pm 10\%$ based partly on a sensitivity
404 analysis presented in Auxiliary Material 1.

405

406 **3.2 Continental Patterns**

407 Regional variations in disturbance rates and *NEP* across the conterminous U.S. reflect
408 harvesting practices and regional climates (Tables 1 and Table A2.1, Figures A2.1, A2.2, A2.3).
409 Forests growing in relatively dry settings (e.g. Rocky Mountain South (RMS)) have low *NEP*,
410 contrasted by high carbon sequestration rates in the Pacific Southwest and Northwest, as well as
411 Southeastern and South Central regions (Table A2.1). The largest rates of disturbance, and the
412 largest sinks of carbon stimulated by forest recovery from recent disturbance (“regrowth sinks”),
413 are in Southeastern (SE), South Central (SC), and Pacific Northwest (PNW) regions. These
414 regional biologically driven sinks do not reflect net biome productivity because recovery
415 trajectories do not include the fate of disturbance-induced carbon removals such as carbon taken
416 offsite to lumber, pulp and paper mills or released promptly on-site by natural and anthropogenic
417 fires (see schematic in Figure 1). This is addressed further in the discussion where we present the
418 forest-to-atmosphere carbon exchange.

419 At the continental scale, the biological recovery sink (*NEP*) is estimated to be $164 \pm 28 \text{ Tg C a}^{-1}$
420 (Table 1), or about $71 \text{ g C m}^{-2} \text{ a}^{-1}$ averaged for the 230 million hectares of forestland represented
421 here. Nearly all (84%) of this *NEP* sink results from net growth of live carbon stocks with only a
422 small fraction shared among soil carbon (6%), litter carbon (2%), and coarse woody debris (8%)
423 stocks (Table 2). Our sample includes 93% of the conterminous U.S. forestland, reported to be
424 250 million hectares [EPA, 2008]. Our analysis did not include the Northern Prairie States region
425 (~6% of total area and ~5% of total carbon) because the effort was originally connected to a
426 Landsat remote sensing analysis whose random sample did not draw Landsat scenes for this
427 region. As verification, our stand-age histograms by region generally correspond well with a

428 similar presentation of the same basic data as recently published by *Pan et al.* [2011]. Comparing
429 to regional statistics of forest area and live biomass reported in *EPA* [2008] we find good
430 correspondence overall (Table 1).

431 The estimated uncertainty arising from forest area, aboveground wood biomass, and
432 conversion of diameter measurements to volume and carbon produced relatively small uncertainty
433 estimates in our biomass and fluxes. This is partly due to the continuous, monotonically
434 increasing, and saturating growth form imposed by the process-model approach. This functional
435 form is more plausible than one that would allow abrupt increases and decreases in aboveground
436 live wood biomass with stand development (i.e. stand age) as are commonly found in the
437 inventory data when arrayed as a chronosequence (e.g. Figure 3, 110- 150 year biomass).

438 Imposing the model's growth form has the effect of filtering out some of the variance inherent in
439 chronosequence trajectories of biomass with stand age. Other uncertainties arising from model
440 structure and assumptions about disturbance severity/type, age, partial cutting, natural wood
441 turnover, and a possible age-related decline in productivity are evaluated by judging the impacts
442 of these factors on model output through sensitivity analyses (see Auxiliary 1, Section 2).

443 We used the 1km forest type map to produce a gridded map of *NEP* and its uncertainty (from
444 variances in FIA data) for the conterminous US (Figure 4). Within each region each forest type
445 considered was assigned the regional estimate of *NEP* for that forest type and region. Regional
446 forest *NEP* sinks range from >25 to 200 g C m⁻² a⁻¹ with eastern and western forests generally
447 ranging from 75 to 100 g C m⁻² a⁻¹. The RMS region is predicted to be uniformly <50 g C m⁻² a⁻¹.
448 The discontinuities conforming to state borders between West Virginia and Virginia and between
449 Washington and Idaho occur because the same forest types in each neighboring region have
450 regionally specific and different growth and disturbance rates.

451 As an independent evaluation of our predicted stocks and fluxes we compared our results with
452 five available studies on chronosequences for forest types in the conterminous US. These studies
453 sometimes do not include estimates of both fluxes and stocks for different aged forests and
454 estimates used various biometric and flux measurement approaches. The small number of sites
455 with available data, variability in the data, and issues of extrapolating fine scale measurements to
456 regional responses do not justify quantitative comparisons and demonstrate the need for more of
457 these types of measurements and for finer scale modeling. The results of these comparisons are
458 shown in Auxiliary 3, Figure A3.1. Agreement varies widely between the comparisons at the
459 different sites/regions.

460

461 **4. Discussion**

462 Comparing estimates of the conterminous U.S. forest *NEP* sink from multiple studies (Table
463 3) reveals a general separation between age-accumulation and stock-change methods. This
464 comparison spans estimates for the 1980s to more recent years (e.g. 2005-2006), but this may be
465 justified because atmospheric inversions seem to indicate a long term mean sink in North
466 America during the '80's and '90's but with large interannual variability [*Baker et al.*, 2006].
467 Four of the six age dependent analyses that seek to represent carbon emissions and sequestration
468 with post-disturbance recovery provide lower estimates of the forest *NEP* sink when compared to
469 the four stock-change analyses, with $82 \text{ g C m}^{-2} \text{ a}^{-1}$ versus $154 \text{ g C m}^{-2} \text{ a}^{-1}$ averaged across their
470 respective studies, or 189 Tg C a^{-1} versus 354 Tg C a^{-1} when integrated across US forest area.
471 This is even true when process-oriented studies rely on forest inventory data to prescribe the rate
472 of aboveground carbon stock recovery with time, as well as the area of forest of different ages.

473 For example, regarding *NEP* alone we find general agreement with *Turner et al.* [1995] who
474 reported 203 Tg C a⁻¹ compared to our estimate of 164 Tg C a⁻¹. In contrast, the *EPA* [2008]
475 stock-change estimate of forest *NEP* is twice as large as this study's age-accumulation result (335
476 compared to 164 Tg C a⁻¹, Table 2). The disparity between the stock-change method and these
477 other, age-accumulation results is likely due to large annual to decadal increases in stocks
478 measured in the inventory that then implies greater *NEP* (regrowth). What causes this general
479 disagreement remains unclear, though growth enhancement is a plausible explanation of the
480 difference, consistent with recent publications [*Cole et al.*, 2010; *Luyssaert et al.*, 2010;
481 *McMahon et al.*, 2010; *Thomas et al.*, 2009]. Effects of growth enhancement are implicit in the
482 stock-change method but not well incorporated in the age-accumulation methods that emphasize
483 effects of regrowth dynamics, even when these methods rely on inventory-derived
484 chronosequences to constrain biomass accumulation as in the present study (see Auxiliary 4 for
485 an illustration of this). There is also one study reported in Table 3 including only the effects of
486 climate and CO₂ fertilization based on an ensemble of models for the conterminous US [*Schimel*
487 *et al.*, 2000]. If this sink were added to the forest recovery (age-yield table) estimates the results
488 would be more in line with the stock change approach.

489 We note that the *EPA* [2008] estimate of total removals is 38% higher than that estimated with
490 our modeling approach (=162/117, Table 3). About half of the difference is due to elevated fire
491 emissions reported in *EPA* [2008], however this estimate is much higher than the rate of forest
492 fire emissions being reported elsewhere (e.g. *van der Werf et al.*[2010]). This difference
493 translates directly into the *NEP* estimated from the stock change method, and elevates the *EPA*
494 [2008] estimate by 20 TgC a⁻¹ relative to the estimate from our approach. The *EPA* [2008] report
495 also estimates 25 TgC a⁻¹ greater removals by harvest. There are two ways we could adjust our

496 methodology to try to match this rate of removal. We could either, a) increase the amount of
497 biomass removed by disturbances on average by increasing the age and hence biomass of
498 disturbed forests, b) increase the amount of biomass removed on average by removing a larger
499 fraction of pre-disturbance biomass and leaving less to decompose on site, or c) increase the area
500 of forests disturbed by increasing the young-aged fraction of forests if we believe the stand age
501 attribute offers a biased representation. All of these would increase removals but they would have
502 different effects on *NEP*. The first option would decrease *NEP* because more disturbance-killed
503 material would be left on-site to decompose and be emitted from forests. The second approach
504 would increase *NEP* because of reduced on-site decomposition. The third approach would
505 decrease *NEP* because a larger fraction of forested area would be concentrated at young stand
506 ages (<15 year old) where *NEP* is either a large negative value or near zero (Figures A2.1, A2.2).
507 And in the extreme case that we simply adjusted our *NEP* estimate upwards to cover the
508 difference in removals, the *EPA* [2008] estimate would still be 126 TgC a⁻¹ higher than the
509 estimate emerging from our age-accumulation method.

510 A term-by-term comparison between stock changes reported from inventory methods and
511 those derived in the current study's age-accumulation approach indicates that a change in live
512 carbon stocks makes up a large portion of the difference in *NEP* estimated with the two methods
513 (Table 2). Annual increases in soil carbon, coarse woody debris, and litter pools are also
514 noticeably lower in the present analysis compared to those reported by the *EPA* [2008; 2010]
515 (Table 3). Because our method, necessarily, produces aboveground live wood biomass and forest
516 area estimates that are consistent with, or wholly derived from, the inventory itself (Table 1), our
517 relatively low estimate of annual changes in live stocks (Table 2) does not appear to be caused by
518 underestimation of a) stocks, or b) forest area. These differences translate to the full forest sector-

519 atmosphere net exchange, whereby the stock-change method estimates a much larger forest sector
520 C sink than obtained with this study's age-accumulation approach (Table 2).

521 Our maps of conterminous US forest *NEP* and its uncertainty (Figure 4) are one of the first of
522 which we are aware (though see *Woodbury et al.* [2007]) and will be used in further study of the
523 impact of the forest disturbance fluxes on atmospheric CO₂ as a boundary flux for atmospheric
524 transport models much as gridded fire, fossil fuel burning, and ocean CO₂ fluxes are prescribed in
525 forward and inverse atmospheric modeling [e.g. *Peters et al.*, 2007]. Complete accounting of
526 forest sector fluxes would additionally require maps of fire [e.g. *van der Werf et al.*, 2010] and
527 wood products emissions. These studies will allow assessment of the detection limits for the
528 magnitude and spatial variability of sinks in top-down studies.

529 This study's approach imposed a number of simplifying assumptions that were necessary
530 given the initial scope of our work. Below we address some of these and their potential
531 implications regarding interpretation of our results.

532 • We assume characteristic regrowth trajectories regardless of disturbance type even though
533 the nature of post-disturbance carbon dynamics is sure to vary between fire, harvest,
534 hurricane, and the severity of disturbance. For instance, around twice as much coarse
535 woody debris (CWD) may remain on site after a severe fire compared to clear-cut harvest
536 [*Tinker and Knight*, 2000]. This remaining detritus provides a source of CO₂ for a
537 prolonged period after disturbance. Using data reported by *Smith et al.* [2009] and the
538 National Interagency Fire Center (to account for Alaskan fires) we estimate that for the year
539 2004 the ratio of burned area to harvested area in the eastern US was about 0.30 compared
540 to 0.46 in the west. In terms of carbon removals though, our forest fire estimates from the
541 Global Fire Emissions Database v3 (10 Tg C a⁻¹) are much smaller than our estimated

542 harvest removals (107 Tg C a⁻¹). Because the total removals are dominated by harvest, as is
543 the total area disturbed, accounting for differences caused by fire versus harvest would not
544 significantly change our results or conclusions. Furthermore, some but not all of this
545 variation is captured by the Monte Carlo approach, as well as with stratification by site
546 productivity and across regions. Partial disturbances such as defoliation events are not
547 represented with the current methodology, and discussed further below.

- 548 • Our assumption of equivalence between forest age and time since disturbance does not
549 account for the effects of partial disturbance that allows older aged trees to remain among
550 regenerating cohorts or the dynamic state of old forests that have reached the age of natural
551 mortality and reestablishment. This particular issue has been examined by *Bradford et al.*
552 [2008] for a subalpine forest system. In that study a large part of the age versus years since
553 disturbance discrepancy arose in stands undisturbed for long periods of time (>200 years),
554 longer than what we analyze in this work. From FIA data we estimate that about 3% of
555 forested land is >200 years old for conterminous US.
- 556 • Our analysis is sensitive to biases in the ages associated with the aboveground live wood
557 biomass trajectories, as explored in an extensive sensitivity analysis described in Auxiliary
558 1, Section 2. For instance, if the FIA ages are older (younger) than actual stand ages, our
559 predicted recovery sink is underestimated (overestimated). This, of course, is an issue with
560 any approach proposing to use FIA age structure information to estimate fluxes and stocks
561 [e.g. *Pan et al.*, 2011]. Despite this sensitivity, we note that bias in stand age is not likely to
562 be large enough to explain the major differences between the stock-change and age-
563 accumulation methods (Table A1.2).

- 564 • The FIA data we used to construct aboveground live wood biomass trajectories include the
565 effects of partial cuts, which are a significant component of disturbance in US forests
566 contributing >50% of the total harvested area [*Smith et al.*, 2009]. Reported stand ages
567 reflect the trees not cut while the plot level biomass will be lower in these cases producing
568 lower regional aboveground live wood biomass for mid and older aged stands. These partial
569 cutting practices (e.g. salvage logging, selective logging, thinning), which remove biomass
570 from forested plots without resetting the FIA-recorded stand age, could have a substantial
571 influence on the forest *NEP* estimate. The implicit inclusion of plots that experienced
572 partial cutting (not fully stocked) likely results in correct biomass estimates but lowers the
573 slope of regrowth trajectories resulting in some underestimation of *NEP*. In an extensive
574 sensitivity analysis (Auxiliary 1, Section 2) we find strong sensitivity to such biases, with a
575 10% elevation of biomass leading to a 14% elevation of conterminous US forest *NEP*. This
576 is equivalent to a 2.3 Tg C a⁻¹ increase in *NEP* for each 1% increase in biomass. Despite
577 this large sensitivity to biomass trajectories, to account for the approximately 160 Tg C a⁻¹
578 difference, the reported biomass would need to have been underestimated by 70% (=160 Tg
579 C a⁻¹ / 2.3 Tg C a⁻¹ per 1% increase in biomass). Additional sensitivity analyses examining
580 effects of natural, partial disturbances that lead to wood turnover and on site decomposition
581 (e.g. ice storms, blowdowns, insect damage) indicate that they are also unlikely to present a
582 large error/bias in our estimate.
- 583 • We do not take into account annual changes in forest area which could contribute to the
584 discrepancy between recovery and stock change approaches. The *EPA* [2008] reports
585 indicate that forest area has been increasing at a rate of 0.24% a⁻¹ since 1990. If we assume
586 that new forests would range between 1 to 5 kg C m⁻² over an age range of 0 to 20 years

587 (e.g. see Figure 3) then the average accumulation rate for these forest would be about 250 g
588 C m⁻² a⁻¹. Correcting this for the increase in forest area produces an added 1.7 Tg C a⁻¹ sink,
589 indistinguishable within the uncertainties of our method.

590 • It has been proposed that forest carbon sinks may be driven by long term trends in
591 temperature, precipitation, nitrogen deposition, and atmospheric CO₂. Responses to these
592 trends are embedded in the biomass-age trajectories from the inventories in complex ways
593 and more recent increases in growth may not be accounted for in our approach (see
594 Auxiliary 4 for a thorough examination of this). Others have addressed this and concluded
595 that forests are not responding in a systematic way to these trends [*Caspersen et al.*, 2000],
596 that forest inventory data are not precise enough to resolve expected responses to trends
597 [*Joos et al.*, 2002], and that a smaller number of inventory measurements on forests of
598 known disturbance history do indeed show strong trends in growth enhancement correlated
599 with trends in temperature and atmospheric CO₂ [*McMahon et al.*, 2010; *Thomas et al.*,
600 2009]. In a study of global terrestrial carbon sinks using CASA, *Thompson et al.* [1996]
601 showed that in order to obtain a terrestrial carbon sink of ~2 Pg C/yr broadly consistent with
602 top-down sink estimates, *NPP* has to undergo a sustained increase of 0.18% per annum.
603 Similar estimates have been reported by others [e.g. *Joos et al.*, 2002]. Our own sensitivity
604 analysis (not shown) showed that a sustained increase in *NPP* of 0.2% per annum would
605 increase live biomass in a typical 60 year old forest by approximately 5% and is thus a weak
606 or undetectable signal in a biomass chronosequence. A 0.2% annual increase in *NPP* is
607 implausibly large sensitivity of photosynthesis to CO₂ ($dNPP/NPP \times CO_2/dCO_2$ of ~0.96, or
608 near proportional response) and would require other positive feedback mechanisms such as
609 nitrogen fertilization and/or climate trends to operate in parallel. We conclude that plausible

610 responses of forest sinks to climate and CO₂ or N cannot be resolved with FIA biomass-age
611 trajectories alone such as those we utilize here and that have been proposed by others [e.g.
612 *Pan et al.*, 2011].

613 The approach described here is also sensitive to uncertain parameters including rates of wood
614 mortality and coarse woody debris decomposition, as well as the amount of dead aboveground
615 and belowground biomass left to decompose onsite following disturbance. It lacks a standing
616 dead wood pool that may be important because it decomposes much more slowly than dead wood
617 in contact with the forest floor [e.g. *Harmon and Hua*, 1991; *Harmon et al.*, 2004; *Janisch et al.*,
618 2005]. In our ongoing efforts, literature is being exhaustively explored to better constrain these
619 and other parameters and processes. Additional effort is being invested in attributing disturbances
620 to particular drivers based on spatial and geospatial records of fire and insect outbreaks. While
621 valuable, it is unlikely that such refinements and constraints will reconcile the large differences
622 between the age-accumulation and stock-change approaches, something that may benefit from a
623 close collaboration with inventory experts to clarify differences of approach and accounting, as
624 well as more comprehensive assessment of possible growth enhancement effects. Future efforts
625 at improving this study's approach will include more detailed prescriptions of type and severity of
626 disturbances, further comparisons with site observations as they become available, and analyses
627 of top-down atmospheric constraints on source/sink magnitude and distributions. Estimates
628 would also be better constrained if additional data on litter, dead wood and soil organic carbon
629 dynamics were available from field studies.

630 **5. Conclusions**

631 Forest Inventory and Analysis data provide unique and valuable information about
632 disturbance history and associated carbon stocks and fluxes with forest recovery. By using these

633 data to constrain forest growth rates in a carbon cycle model, this study provides a more detailed
634 estimate of carbon sources and sinks from recent forest disturbance and recovery across regions
635 and forest types of the US. One of our key findings is a much smaller net sink of carbon in
636 conterminous US forests than previously estimated with the stock-change approach as used in
637 UNFCCC reporting [EPA, 2008]. The source of across study inconsistencies among national
638 estimates of stocks and fluxes remains largely unexplained. The paucity of observed net
639 ecosystem productivity and biomass chronosequences limits our ability to evaluate modeled
640 responses. These types of observations are critically needed in order to adequately test models
641 representing disturbance and subsequent recovery.

642

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831 **Figure Captions.**

832 **Figure 1.** Schematic diagram illustrating stock and flux (italicized) relationships between the
833 forest sector and atmosphere. The entire forest sector net flux (sink) as defined by the stock-
834 change approach is: $\text{Net Flux} = \Delta C_{\text{stocks}} + \Delta C_{\text{wood products}}$. Alternatively, using our model driven
835 estimates of *NEP* it is: $\text{Net Flux} = \text{NEP} - \text{Wood Products Emissions} - \text{Fire}$.

836 **Figure 2.** Conterminous U.S. distribution of forest type groups shown with thick state boundaries
837 that trace regions from the Resource Planning Act Assessment by the US Forest Service. Colors
838 differentiate FIA forest type groups. The rectangles represent areas where gridded climate and
839 phenology were used in the simulation of fluxes and stocks for each forest type within each
840 rectangle.

841 **Figure 3.** Characteristic trajectories of aboveground live wood biomass regrowth and associated
842 carbon sources / sinks (expressed as net ecosystem productivity, *NEP*) following a stand-
843 replacing disturbance in high productivity Douglas-fir stands of the Pacific Northwest. Results
844 are from the CASA model fit to regrow stocks consistent with 25 independent samples from the
845 forest inventory data (red circles). Net releases in the year following disturbance are as low as -
846 $3000 \text{ g C m}^{-2} \text{ a}^{-1}$ (see Auxiliary Material 2, Figure A2.2) rising to above $-500 \text{ g C m}^{-2} \text{ a}^{-1}$ in the
847 second year of regrowth.

848 **Figure 4a,b.** Map of average net ecosystem productivity (top, a) and uncertainty expressed as
849 one standard deviation (bottom, b) (*NEP* in $\text{g C m}^{-2} \text{ a}^{-1}$) for forests of the conterminous US.

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852 Table 1. Regional distribution of forest area, live biomass (Live B), ratio of EPA [2008] to this
 853 study's forest area (f_{EPA08} Area), ratio of EPA [2008] to this study's live biomass (f_{EPA08} Live B),
 854 net ecosystem productivity (NEP), fraction of forest that is less than 25 years old (<25y), less than
 855 5 years old (<5y).

| Region | Area | Live B | f_{EPA08} | f_{EPA08} | NEP | <25y | <5y |
|------------|---------------------------|----------|-------------|---------------|-------------------------|------|-----|
| -- | [10^9 m ²] | [Tg C] | Area [--] | Live B [--] | [Tg C a ⁻¹] | [%] | [%] |
| NE | 339 | 3,253 | 1.11 | 1.01 | 32±5.5 | 10 | 2 |
| NLS | 212 | 1,236 | 0.99 | 1.11 | 12±1.3 | 16 | 3 |
| SE | 355 | 2,621 | 1.00 | 0.94 | 30±3.5 | 39 | 8 |
| SC | 384 | 3,220 | 1.27 | 1.00 | 40±4.2 | 37 | 8 |
| RMN | 192 | 1,189 | 0.98 | 1.10 | 7±1.8 | 21 | 5 |
| RMS | 493 | 1,815 | 0.81 | 0.97 | 11±5.5 | 1 | 0 |
| PSW | 127 | 1,522 | 1.06 | 0.95 | 13±2.8 | 11 | 2 |
| PNW | 202 | 2,162 | 1.05 | 1.13 | 18±3.0 | 19 | 4 |
| Total/Mean | 2,303 | 17,017 | 1.03 | 1.08 | 164±27.7 | 17 | 4 |

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858 Table 2. Changes in carbon stocks [Tg C a⁻¹] in the year 2005 reported in different studies.
 859 Italicized values are inferred from mass balance.

| | This Study | <i>EPA</i> [2008] |
|--|------------|-------------------|
| ΔTotal Soil C | 3 | 9 |
| ΔLitter C | 1 | 15 |
| ΔCoarse Woody Debris (CWD) | 4 | 16 |
| ΔCWD Below | 0 | -- |
| ΔLive C | 39 | 133 |
| Total Stock Change | 47 | 173 |
| Removals ^g | 117 | <i>162</i> |
| Harvest ^a | <i>107</i> | 132 |
| Wildfire Emissions ^b | 10 | 30 |
| <i>NEP</i> ^c | 164 | 335 |
| Wood Products Emissions ^d | 102 | 102 |
| Wood Products Storage ^e | 5 | 30 |
| Forest Sector - Atmosphere Exchange ^f | 52 | 203 |

860 ^athis study inferred as: Harvest = Removals – Wildfire Emissions;

861 ^bthis study estimated wildfire emissions from the Global Fire Emissions Database v3 (GFED3)
 862 [*van der Werf et al.*, 2010];

863 ^cfor the purposes of this table calculated as:

864 $NEP = \Delta\text{Total Soil C} + \Delta\text{Litter C} + \Delta\text{CWD} + \Delta\text{CWD Below} + \Delta\text{Live C} + \text{Removals}$; values
 865 differ from those in Table 1 due to differences in the method of aggregation and associated
 866 averaging of terms;

867 ^dthis study adopted values reported in *EPA* [2008];

868 ^ethis study calculated as:

869 Wood Products Storage = Removals – Wood Products Emissions – Wildfire Emissions;

870 ^fthis study calculated as:

871 Forest Sector-Atmosphere Exchange = *NEP* – Wood Products Emissions – Wildfire Emissions;

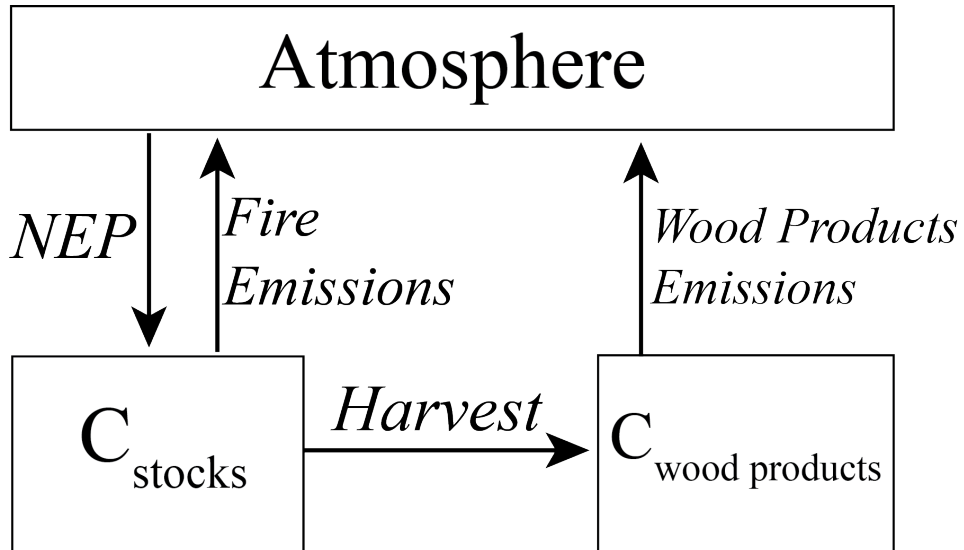
872 ^gfor *EPA* [2008] calculated as: Removals = Wildfire Emissions + Harvest.

873 Table 3. Forest carbon *NEP* and stock change for the conterminous US [Tg C a^{-1}] from this and a
 874 sample of previously published estimates. Estimates are classified according to approach: age
 875 structure–C accumulation (*A&A*), stock change ($\Delta C_{stocks}=NBP$), or process model (*P*), where *P* is
 876 a process model ensemble result that accounts for CO_2 and climate effects [*Schimel et al.*, 2000],
 877 and *Pacala et al.* [2001] combines approaches for an overall estimate and range. Low and High
 878 refers to 1 standard deviation about the mean estimate.

| Source | Approach | Mean | | Low | High | ΔC_{stocks} | Harvest | Fire |
|--|---------------------|------------|-----|-----|------|---------------------|---------|------|
| | | <i>NEP</i> | | | | | | |
| <i>Schimel et al.</i> [2000] ^a | <i>P</i> | 80 | | | | | | |
| This Study ^b | <i>A&A</i> | 164 | 136 | 192 | 47 | 107 | 10 | |
| <i>Houghton et al.</i> [1999] ^c | <i>A&A</i> | 182 | | | 10 | 92 | 80 | |
| <i>Turner</i> [1995] ^c | <i>A&A</i> | 203 | | | 79 | 124 | 0 | |
| <i>Houghton</i> [2003] ^d | <i>A&A</i> | 207 | | | 35 | 92 | 80 | |
| <i>Woodbury et al.</i> [2007] ^f | ΔC_{stocks} | 270 | 256 | 293 | 108 | 132 | 30 | |
| <i>EPA</i> [2008] ^g | ΔC_{stocks} | 335 | | | 173 | 132 | 30 | |
| <i>Birdsey & Heath</i> [1995] ^h | ΔC_{stocks} | 368 | | | 211 | 127 | 30 | |
| <i>Hurt et al.</i> [2002] ⁱ | <i>A&A</i> | 372 | 282 | 442 | 230 | 92 | 50 | |
| <i>Pacala et al.</i> [2001] ^j | synthesis | 392 | 312 | 472 | 220 | 92 | 80 | |
| <i>King et al.</i> [2007] ^k | ΔC_{stocks} | 411 | 383 | 439 | 236 | 145 | 30 | |

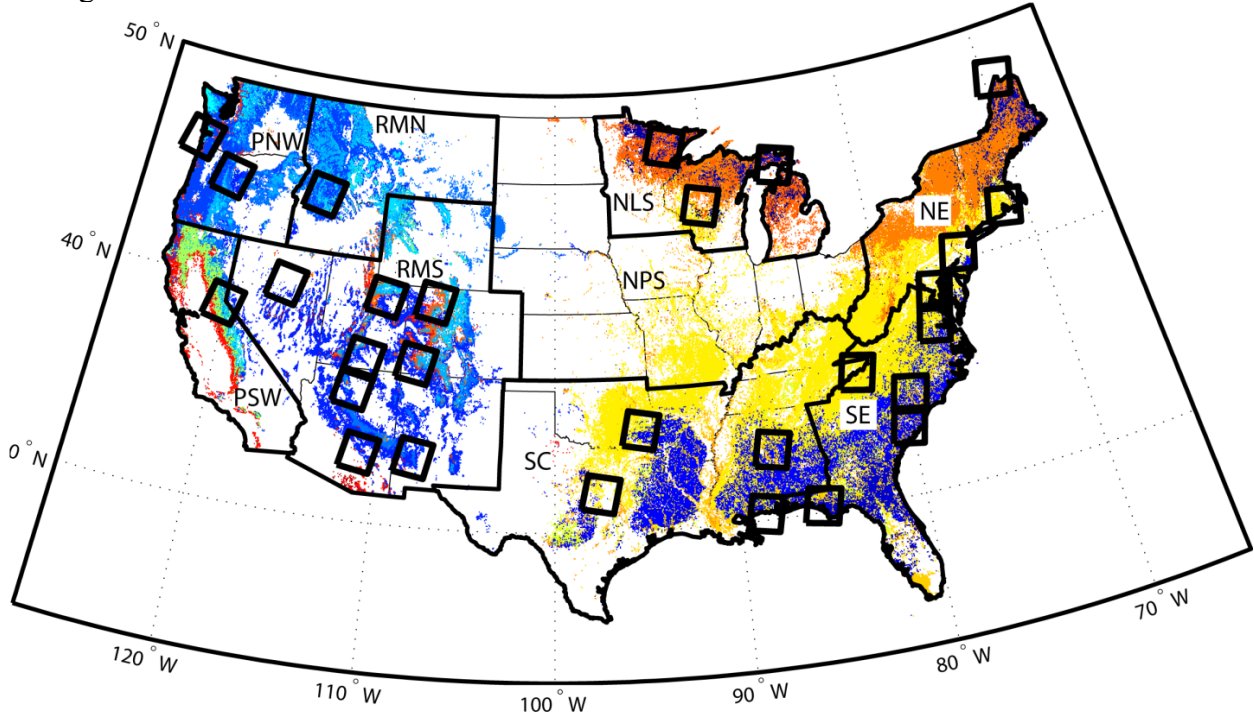
- 879 a) for 1980-1993
 880 b) for 2005, C stock change = $NEP - Harvest - Fire$, see Table 2, our total removals are 117
 881 Tg C a^{-1} that includes fire and harvest, assume fire at 10 Tg C a^{-1} (see GFED3 of *van der*
 882 *Werf et al.* [2010] and *Zheng et al.* [2011])
 883 c) 1980s
 884 d) for 1990's, harvest and fire from *Houghton et al.* [1999]
 885 e) for ~1990,
 886 f) for 2005
 887 g) for 2005
 888 h) for 1992
 889 i) for 1980s
 890 j) for 1980's
 891 k) for 1980s
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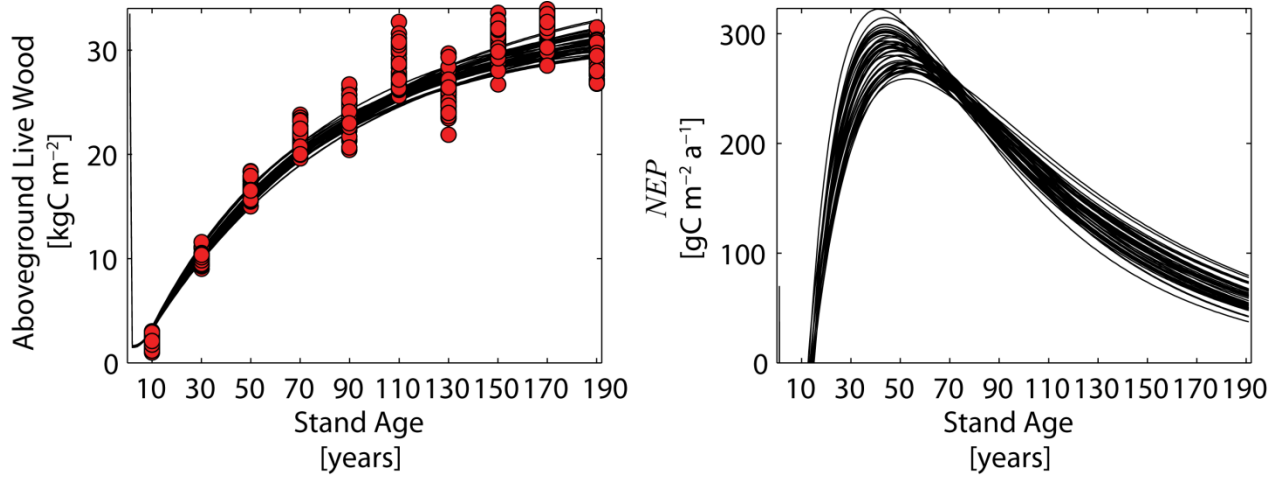
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