

1 **Technical Note: Linking climate change and downed woody** 2 **debris decomposition across forests of the eastern United** 3 **States**

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12 13 **1 Introduction**

14 Live and dead trees are thought to contain ~60% of forest C in mature forests, while soil and
15 litter pools contribute the remaining ~40% (Ryan et al., 2010). In particular, downed woody
16 debris (DWD) is an important component of overall C stocks, accounting for approximately
17 20% of total C in old-growth (Harmon et al., 1990) and secondary (Bradford et al., 2009)
18 forests. Whether measured in terms of decay rates, C flux, and/or residence time (T_{RES} ; the
19 number of years until a DWD piece loses structural integrity and transitions to another
20 ecosystem pool), previous investigations have quantified DWD C depletion in many regions
21 and forest types (Fraver et al., 2013; Mackensen et al., 2003; Radtke et al., 2009; Russell et
22 al., 2014). The potential for altered decomposition is well recognized for C stored in soil
23 (Conant et al., 2011; Davidson and Janssens, 2006; Giardina and Ryan, 2000) and litter
24 (Brovkin et al., 2012; Prescott, 2010), however, the impact of changing environmental
25 conditions on DWD debris dynamics remains largely unexplored. While it has been shown
26 that fungal colonization and termite biomass account for nearly three-quarters of the
27 variability in wood decomposition at the local scale (Bradford et al., 2014), the factors driving
28 variability at regional scales remain largely unexplored, particularly under changing
29 environmental conditions. It is essential to understand and quantify these temporal patterns as

1 DWD represents not only a substantial C pool but also facilitates tree regeneration, is a
2 determinant of fire behavior, and serves as vital wildlife habitat whose dynamics may be
3 altered under future climates.

4 Although DWD decomposition is included in a variety of ecosystem simulation models
5 primarily through their relationship with temperature (e.g., Kirschbaum, 1999; White et al.,
6 2000), the degree to which DWD decomposition may be altered under potential global change
7 scenarios remains to be quantified or incorporated into projections of long-term forest C
8 dynamics. One unknown is the influence that projected future climates may have on DWD
9 decomposition rates. The importance of temperature and moisture as drivers of DWD
10 decomposition is well established (Edmonds et al., 1986), underscoring the potential for
11 climate change to alter the future dynamics of this critical ecosystem component. Increased
12 rates of decomposition will likely reduce the duration that woody debris is available for dead
13 wood-dependent organisms (Mazziotta et al., 2014). Woody detritus decomposition rates
14 depend not only on climate but also; DWD size and condition, the decomposer community,
15 and geographic locale (Bradford et al., 2014; Brovkin et al., 2012; Fraver et al., 2013; Radtke
16 et al., 2009; Russell et al., 2013; Russell et al., 2014). Specifically, the eastern United States
17 has experienced increased mean annual temperatures (MAT) throughout much of the region
18 over the past century with the exception of the Southeast, which is characterized by both
19 cooling winters and warmer summers (Zhu et al., 2012). Woody debris decomposition rates
20 should be sensitive to these changes in MAT (Brovkin et al., 2012), recognizing that local-
21 scale factors additionally contribute to determining wood decay patterns (Bradford et al.,
22 2014). To accurately represent DWD dynamics in ecosystem processes, models should be
23 sensitive to transient responses, such as changes in disturbance regimes, when depicting the
24 nonlinear patterns inherent in DWD decomposition (Harmon et al., 2011a).

25 Our objective was to link current and future climate information with models representing
26 woody debris decomposition to quantify potential future changes in temporal DWD dynamics
27 in forests of the eastern US. Specific objectives were to (1) compare differences in DWD
28 residence time assuming a static versus dynamic climate throughout the duration of
29 decomposition and (2) forecast ecosystem-level C flux for DWD using the static and dynamic
30 climate scenarios.

31

1 2 Methods

2 2.1 Study area

3 The geographic scope investigated here ranged eastward from the US state of Minnesota to
4 Maine in the north and Louisiana and Georgia in the south (latitude range from 29.56°N to
5 48.74°N; longitude range from 67.06°W to 96.71°W). Data from the eastern US are employed
6 because remeasurement data have been collected in the region which have informed DWD
7 modeling efforts (e.g., Russell et al., 2013), whereas remeasurement data are not yet available
8 for western US states. Forest types across this region varied in terms of species assemblage,
9 forest productivity potential, and climate. More than 75 forest types were identified by the US
10 Department of Agriculture Forest Service's Forest Inventory and Analysis (FIA) program
11 across the study area, which represented 14 broader forest type groups (Woudenberg et al.,
12 2010).

13 2.2 Data

14 Data used to simulate the decomposition of woody debris were obtained from a DWD
15 inventory conducted in 2001 on 516 FIA plots across the eastern US (Russell et al., 2014).
16 Each plot consisted of four 7.32-m fixed radius subplots for a total plot area of approximately
17 0.07 ha where tree and site attributes were measured. Downed woody pieces were defined as
18 DWD in forested conditions with a diameter greater than 7.62 cm along a length of at least
19 0.91 m. All plots displayed a minimum of at least one DWD piece that met this definition.
20 Individual DWD pieces were sampled using a line-intercept sampling method (Van Wagner,
21 1968) on 18.0-m horizontal distance transects radiating from each FIA subplot center at
22 azimuths of 30, 150, or 270 degrees. Only two transects from the three azimuths were
23 sampled within each subplot depending on spatial arrangement (30° and 150° for north and
24 southeast subplots; 150° and 270° for center and southwest subplots), totaling 143.6 m for an
25 entire inventory plot. Data collected on each DWD piece included small-end and large-end
26 diameters, decay class (DC), length (LEN), species, and piece location (i.e., plot, subplot, and
27 transect number; horizontal distance along a sampling transect). In the field, DC was assigned
28 to each DWD piece using a five-class system, with 1 being least and 5 being most decayed.
29 Piece LEN was defined as the total length of the log in m. In total, 4,384 DWD pieces were
30 collected from 32 conifer and 87 hardwood species as part of the inventory.

1 Climate information was obtained by specifying latitude, longitude, and elevation of each FIA
2 plot location to a spline surface model developed from climate station data across forests of
3 North America (Table 1; Fig. 1) (Rehfeldt, 2006; US Forest Service, 2014). Downscaled
4 predictions from 17 CMIP5 model outputs were used to assess differences in DWD
5 decomposition rates resulting from future climate predictions: RCP 4.5, RCP 6.0, and RCP
6 8.5 (Intergovernmental Panel on Climate Change (IPCC), 2013). Downscaled GCM data were
7 obtained from the Moscow (Idaho) Forestry Sciences Laboratory
8 (<http://forest.moscowfsl.wsu.edu/climate>), which were produced by adapting spline surfaces
9 from present climate data to GCM predictions (Rehfeldt, 2006).

10 **2.3 Analyses**

11 Decay class transition models were used to project the mass loss of DWD (Russell et al.,
12 2013). Here, DC transition was defined as the probability that a DWD piece would remain in
13 the same DC or advance to subsequent DCs at repeated inventories. These DC transitions
14 were estimated as a function of climate (as measured in the number of degree days greater
15 than 5° C [DD5] observed on the FIA plot), LEN, and DC (Russell et al., 2013). [Given the
16 relationship between log size and woody debris decomposition \(Cornwell et al., 2009; Janisch
17 et al., 2005; Mackensen et al., 2003\), LEN was used as a predictor in the DC transition model.](#)

18 We assumed a conic-paraboloid form (Fraver et al., 2007) to estimate the initial volume (*Vol*)
19 of each DWD piece. Initial density (*ID*; kg m⁻³) for an individual species *m* (Harmon et al.,
20 2008) and the appropriate DC reduction factor (*DCRF*) for DWD of a given species group *n*
21 in a DC *k* (Harmon et al., 2011b) was obtained to estimate losses in wood density. To
22 accurately represent DWD mass loss, a volume reduction factor (*VERF*) was subsequently
23 applied to account for structural reductions in DWD *Vol* as decay progresses. We applied a
24 *VERF* of 1,1,1, 0.800, and 0.412 for DC 1, 2, 3, 4 and 5 pieces, respectively (Fraver et al.,
25 2013). Hence, DWD mass was calculated as:

$$26 \text{ Mass} = ID_m * DCRF_{kn} * Vol * VERF_k \quad (1)$$

27 where all variables are as previously defined.

28 A Monte Carlo simulation framework was used to estimate DWD *Mass* in five-year intervals
29 using the DC transition equations (Russell et al., 2013; Russell et al., 2014). For the
30 simulations, 1,000 runs were carried out for 200 years to introduce uncertainty in estimating
31 DC changes. This method involved simulating the DWD pieces by first assuming they were

1 non-decayed, then drawing a random number from a uniform distribution and comparing it to
2 the cumulative five-year probability predicted using the DC transition model. Downed woody
3 debris DC transitions were estimated by predicting the cumulative probabilities of pieces
4 advancing in decay using a cumulative link mixed model. Cumulative link models (CLMs)
5 are a type of ordinal regression model in which response variables are considered categorical
6 or ordered (Agresti, 2007). The variables DD5, LEN, and initial DC, were used to indicate
7 decomposition potential across the eastern US and thus estimate DWD DC transitions
8 (Russell et al., 2014):

$$9 \text{ logit}(\gamma_{ikj}) = \theta_k - \beta_1 \text{DD5} - \beta_2 \text{LEN} - u_{\text{ForType } j} + \varepsilon \quad (2)$$

10 where θ_k is the intercept term for DC k (i.e., DC 1, DC 2, DC 3, DC 4, or DC 5), γ is the
11 cumulative probability for DWD piece i moving through each of the successive k decay
12 classes within each ForType j , β_i are the parameters estimated for conifer and hardwood
13 species separately, and ε is the random residual term. The random effect u was specified to
14 represent forest type-specific effects on the DC transition process. The finding that LEN was
15 a more effective predictor of decomposition than log diameter in these DC transition models
16 is consistent with other studies that suggested a lack of a consistent relationship between log
17 diameter and woody debris decomposition (e.g., Harmon et al., 1987; Radtke et al., 2009).

18 Predictions were accomplished by applying the DWD DC transition equations (Russell et al.,
19 2013) to the data described above using the simulation framework. For each of the 4,384
20 DWD pieces, a 1,000-run Monte Carlo simulation was performed up to 200 years.

21 We assumed DWD would decay according to one of two scenarios: (1) a fixed (i.e., static)
22 climate throughout the timespan of DWD decomposition, or (2) a dynamic climate throughout
23 DWD decomposition depending on the future climate predicted at each FIA plot location.

24

25 **2.4 DWD decomposition scenarios**

26 **2.4.1 Baseline**

27 For a baseline scenario, a static climate was assumed throughout the timespan of DWD
28 decomposition. Hence, the independent variable DD5 used to represent climate regime in the
29 DWD DC equation was held fixed and assumed to be the thirty-year (1961-1990) normal

1 depending on its location within the region. To compare relative differences in DWD
2 decomposition, pieces were separated into species group (i.e., conifer and hardwood species)
3 and geographic region (i.e., north and south; Supplementary Table S1). Smaller sample sizes
4 for some species (e.g., <10 DWD pieces) constrained us to analyze relative differences
5 according to the general species group.

6 Simulating the DWD pieces allowed us to approximate the number of years in which the
7 proportion of biomass remaining attained any specified proportion. Residence time (T_{RES}) was
8 calculated as the number of years in which the mean proportion of biomass remaining fell
9 within one standard error of the mean for a decay class 5 log (Russell et al., 2014). From a
10 biological perspective, T_{RES} might be used as a surrogate for the number of years until a DWD
11 piece loses all structural integrity and transitions to another population (i.e., another C pool).
12 At this point, the DWD piece may be incorporated into the soil organic horizon and thus no
13 longer meets the criteria for being inventoried as DWD within the FIA protocol (exclusive of
14 combustion or harvest removal).

15 **2.4.2 Future climate**

16 For a changing climate scenario, a dynamic climate was assumed to occur throughout DWD
17 decomposition. Current CMIP5 models (Taylor et al., 2012) as described in the fifth
18 assessment report (AR5) of the IPCC (2013) were obtained using three scenarios (RCP 4.5,
19 RCP 6.0, RCP 8.5; US Forest Service, 2014). An ensemble of 17 AR5 model predictions was
20 used for each RCP scenario (Supplementary Table S2). Given that the DC transition equation
21 operated on a five-year interval while climate information were provided for the thirty-year
22 normal (1961-1990) and years 2030, 2060, and 2090, values for the DD5 variable were
23 assumed to transition linearly between 2001 and 2030, 2030 and 2060, 2060 and 2090, and
24 post-2090 (if T_{RES} was not yet reached by the year 2090). Within the simulation, a dynamic
25 DD5 variable resulted in different values for T_{RES} and C flux when compared to the baseline
26 scenario.

27 Projected changes in temperature (i.e., DD5) were more apparent at these sites compared to
28 variables representing moisture such as mean annual precipitation (MAP). Comparing the
29 thirty-year normal with projected 2090 climate, DD5 would increase on average by 39.1%
30 (SD=10.8%) while MAP is projected to increase by only 7.2 cm or 7.1% (SD=2.8%).
31 Regionally, increases in the percent difference in current versus projected DD5 would range

1 from as low as 29.3% (SD=5.4%) in the Southeast to as high as 51.2% (4.5%) in the Northern
2 Lake States (Fig. 1). Hence, in the absence of local-scale factors to use as a surrogate for
3 decomposition (e.g., Bradford et al., 2014), employing temperature differences under future
4 climate scenarios may be used to at least in part to explain DWD flux across the eastern US.

5 **2.4.3 C flux**

6 To scale our estimates of T_{RES} changes for DWD pieces, we forecasted ecosystem-level DWD
7 C flux. This was accomplished by projecting current DWD stocks inventoried from 2007-
8 2011 (hereafter termed “year 2010”) by the FIA program in 29 eastern US states (Woodall et
9 al., 2013). These data were collected in a similar manner to the 2001 data, with the primary
10 difference being that DWD were sampled along three 7.32-m transects at each of four
11 subplots, totaling 87.8 m for a complete FIA plot (Woodall and Monleon, 2008).

12 Current DWD C stocks were first estimated by multiplying plot-level biomass values by a C
13 concentration constant of 0.5 (Mg/ha), followed by a simulation of DWD pieces. Carbon
14 stocks in the DWD pool were then estimated in 5-year time steps from 2010 onward.
15 Assuming no inputs into the DWD pools over a 100-year span, C flux was defined as the
16 amount of C lost for each 5-year span (Mg/ha/5-yr). If the estimate of T_{RES} for a given species
17 was exceeded by the number of simulation years, then it was assumed that the piece had
18 completely decomposed (i.e., biomass was set equal to zero). Means for C flux were
19 summarized by general forest type group (i.e., conifer and hardwood) following multiple
20 simulation runs.

22 **3 Results**

23 Baseline estimates of T_{RES} ranged from 49.9 ± 7.5 years (mean \pm SD) for conifer species in
24 the southern US to 87.4 ± 13.0 years for conifer species in the northern regions (Table 2). For
25 all RCP scenarios, T_{RES} was predicted to decrease for all species groups and regions compared
26 to the baseline scenario (see Table 3 for RCP 6.0). Decreases in T_{RES} were generally less than
27 ten years for southern species, while northern species displayed greater decreases. The
28 decrease in T_{RES} for smaller DWD pieces generally was less than ten years. However, in some
29 cases the decrease in T_{RES} exceeded 20 years for larger DWD (>20 m in length) pieces located
30 in the north (Table 3).

1 We estimated that the mean decrease in T_{RES} was greatest for northern hardwood species.
2 When averaged across all climate models, the maximum mean decrease for this group was
3 10.3 ± 3.5 years, or a decrease of 13%. Decreases in T_{RES} were lowest for both southern
4 conifer species where a 6% decrease was found, followed by northern conifer and southern
5 hardwood species (10%; Fig. 2).

6 Carbon flux was initially greater for RCP scenarios compared to the baseline scenario (Fig.
7 3). For conifer forest types during the first five years, flux ranged from -0.23 ± 0.05 Mg C/ha
8 when considering an RCP 6.0 scenario to -0.26 ± 0.05 Mg C/ha considering an RCP 8.5
9 scenario. Similarly, flux ranged from -0.50 ± 0.10 Mg C/ha when considering the baseline
10 scenario to -0.56 ± 0.08 Mg C/ha for an RCP 8.0 scenario in hardwood forest types during the
11 first five years. Carbon flux generally tended to decrease more rapidly throughout the duration
12 of the simulation (e.g., from 2015 to 2095) for RCP scenarios when compared to that of the
13 static baseline climate assumption.

14

15 **4 Discussion**

16 Our study suggests that increased decomposition rates as resulting from future climate
17 changes will decrease DWD residence times and increase initial C emissions from decaying
18 logs. These findings have direct implications for modeling C dynamics from DWD under
19 future global change scenarios and suggest that future forest management and conservation
20 activities may need to proactively manage for DWD to maintain contemporary levels. Given
21 the range in climate and total number of species, the eastern US was an appropriate region to
22 explore changes in DWD dynamics under future projected climates.

23 Findings of a shorter residence time for northern hardwoods as opposed to conifers assuming
24 a baseline scenario was expected given our general understanding of species differences in
25 wood decay (Cornwell et al., 2009). The observation of the largest percent difference in
26 residence time change when comparing the RCP 6.0 scenario with that of the baseline for
27 northern hardwoods (13%) may be due to greater projected increases in DD5 for the northern
28 compared to southern regions (Fig. 1). The length of DWD pieces will likely further influence
29 DWD residence time if one is interested in a particular species of a general size class (Russell
30 et al., 2014).

1 Future work merging our results with ecosystem models representing tree growth and
2 mortality in conjunction with DWD dynamics could allow for an array of C flux and stock
3 projections (Mazziotta et al., 2014). Moreover, the long-recognized ecological importance of
4 DWD argues for increased empirical and modeling studies that account for the impacts of
5 climate change on this critical component of forest ecosystem functioning (Krajick, 2001;
6 Stokland et al., 2012). Results highlight the need for detailed inventories of DWD so that the
7 stocking in various pools can be assessed with more accurate quantification of decomposition
8 pathways. Future investigations on DWD decomposition rates should focus on employing
9 climate-related parameters in addition to assessments of locale-scale factors (e.g., fungal
10 colonization; Bradford et al., 2014) to examine the response of DWD to potential interactions
11 between altered disturbances and changing climate conditions. Determining how to better
12 incorporate site-specific factors within ecosystem simulations will encourage modelers to
13 investigate the role of local-scale factors in addition to climate for representing DWD
14 decomposition.

15 We note that these simulations did not account for future DWD inputs—we quantified
16 decomposition trajectories of current DWD C stocks under alternative climate scenarios to
17 characterize temperature effects on DWD dynamics independent of other processes.
18 Particularly when examining C flux, incorporating the contribution of live tree C
19 simultaneously with DWD dynamics will better depict total ecosystem C response to changes
20 in climate. Such an approach was recently highlighted by Mazziotta et al. (2014) through their
21 use of a gap-based forest simulation model to forecast changing DWD populations. Given that
22 model parameters for decomposition are largely dependent on temperature in dynamic global
23 vegetation (Cramer et al., 2001), process (Kirschbaum, 1999; Kirschbaum and Paul, 2002),
24 and empirical models that represent DWD decomposition (Crookston et al., 2010; Rebaïn et
25 al., 2010), there is a need to examine the influence of changing temperatures on woody debris
26 dynamics. A key modeling development would be the incorporation of key forest
27 disturbances common to a region (e.g., windstorms, insect and disease outbreaks) in a
28 stochastic framework given the linkage with inputs into the standing and DWD pools.

29 Despite not including C inputs to the DWD pool in this study, emerging research from the
30 same study area suggests that climate change may increase the rate of forest development
31 (i.e., turnover; Zhu et al., 2014). The potentially increased rates of stand development appear
32 to align with our study's projections of increased detrital C emission and hence elevated

1 DWD turnover. The combination of these two results suggest that the residence time of C in
2 the major forest ecosystem pools of live and dead biomass will decrease. Although the effect
3 of decreased residence times on the overall sink strength of forest ecosystems will be
4 dependent on future biomass production rates (e.g., longer growing seasons, droughts, and/or
5 CO₂ enrichment), it does suggest that managers will have less time to consider management
6 options (Malmshemer et al., 2008) as forest biomass becomes more transitory. Moreover,
7 given the critical role of DWD as habitat for a myriad of dead wood-dependent organisms,
8 these future dynamics need to be considered in species vulnerability assessments and action
9 plans, particularly for species requiring habitat elements as refugia during drought and
10 temperature extremes (Amaranthus et al., 1989). Such future dynamics argue for an increasing
11 emphasis on the deliberate retention and creation of DWD habitats in managed landscapes to
12 compensate for accelerated rates of depletion associated with future climate conditions.

13

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1 Table 1. Current climate conditions for 516 plot locations using the US Forest Service
 2 Moscow Laboratory climate model (<http://forest.moscowfsl.wsu.edu/climate/>) for
 3 determining differences in downed woody debris decomposition dynamics across the eastern
 4 US.

Variable	Definition	Units	Mean	SD	Min	Max
DD5	Annual degree days	>5 °C	2667.6	915.2	406.0	5669.0
MAP	Mean annual precipitation	mm	869.8	360.2	219.0	3282.0
MAT	Mean annual temperature	°C	9.2	4.3	-0.3	20.6
MTCM	Mean temperature in the coldest month	°C	-5.3	6.6	-18.0	12.7
MTWM	Mean temperature in the warmest month	°C	22.5	3.2	9.3	28.9

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1 Table 2. Baseline estimates of downed woody debris residence times assuming a static
2 climate scenario.

Species group	Region	<i>n</i>	Residence time (years)	
			Mean	SD
Conifers	North	1648	87.4	13.0
	South	490	49.9	7.5
Hardwoods	North	1581	80.0	16.4
	South	665	51.6	11.0

3 Baseline estimates assume a static climate scenario throughout the duration of decomposition,
4 assuming to be the thirty-year (1961-1990) normal depending on the number of degree days
5 (DD5) >5° C for each plot location *n*, number of observations

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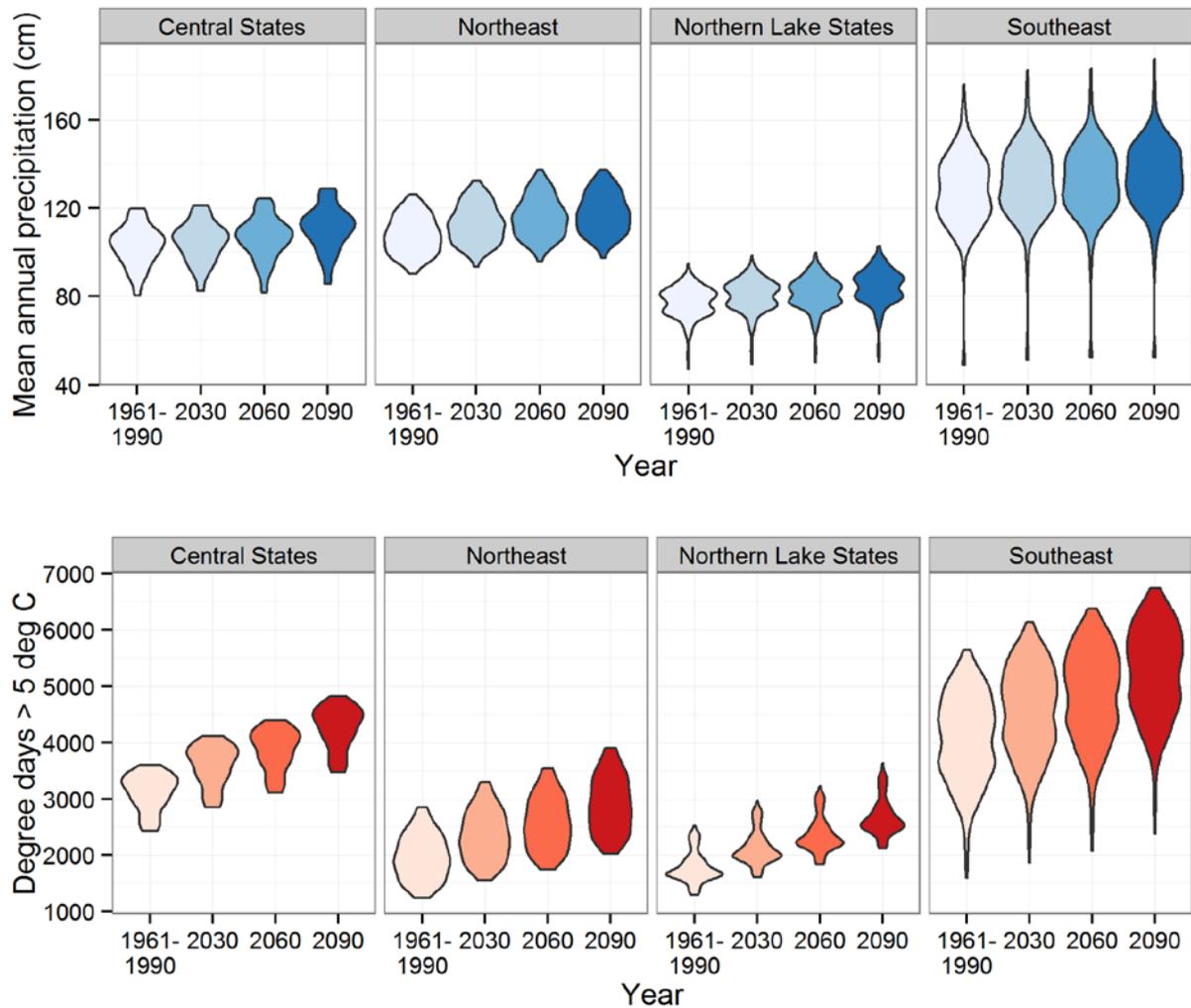
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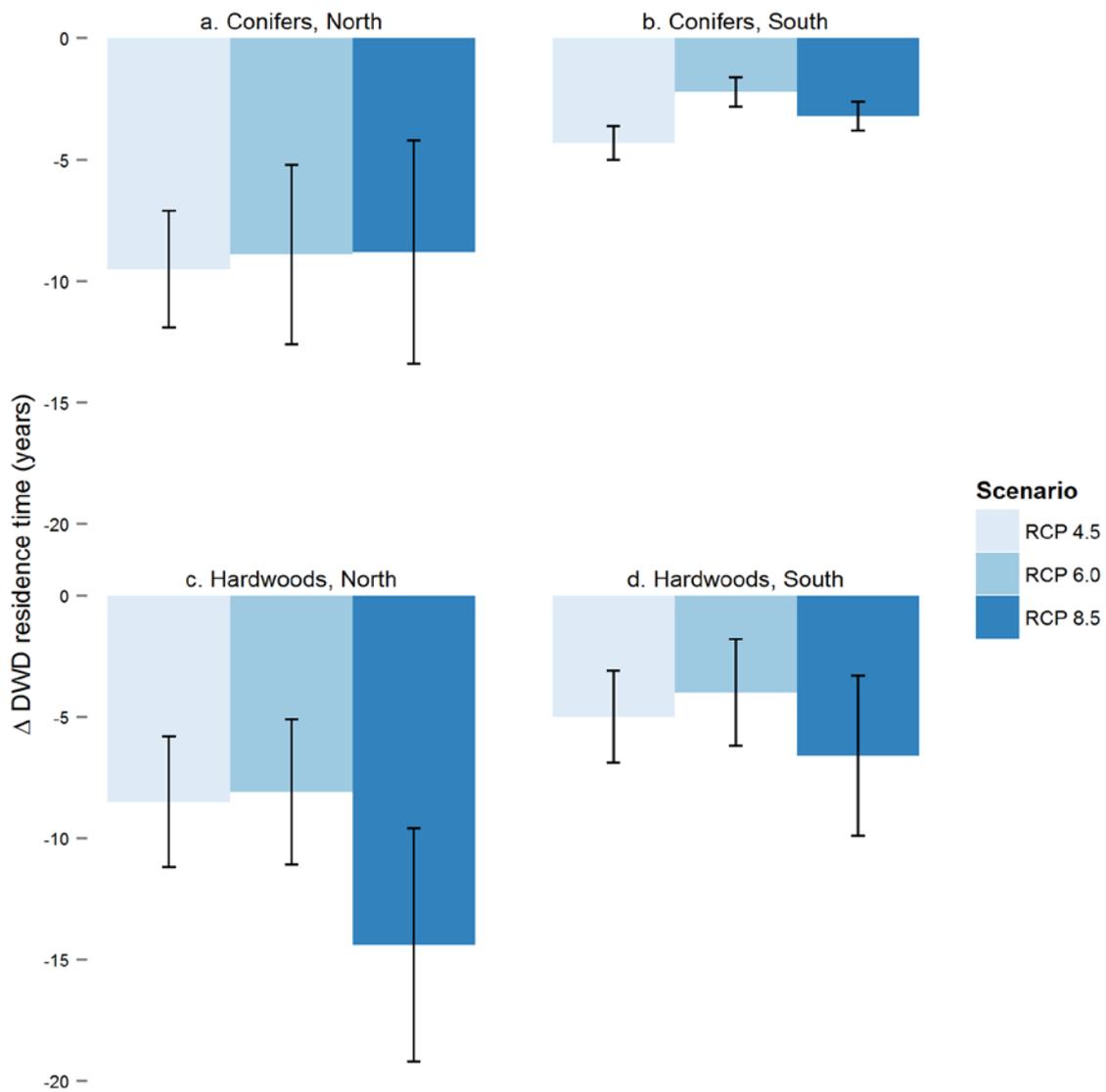
1 Table 3. Distribution of estimated decreases in downed woody debris residence time (years)
 2 by piece size, species group, and region for eastern US forests for a baseline current climate
 3 scenario and assuming changes in future climate for up to 200 years (based on an RCP 6.0
 4 scenario from an ensemble of 17 CMIP5 models).

<u>Species group</u>	<u>Region</u>	<u>Length</u>	<u>Quantile (years)</u>				
			<u>Min</u>	<u>25</u>	<u>50</u>	<u>75</u>	<u>Max</u>
<u>Conifers</u>	<u>North</u>	<u>Short</u>	<u>-9.8</u>	<u>-7.4</u>	<u>-4.9</u>	<u>-4.0</u>	<u>-0.9</u>
		<u>Med</u>	<u>-11.4</u>	<u>-9.9</u>	<u>-9.4</u>	<u>-8.6</u>	<u>-1.2</u>
		<u>Long</u>	<u>-24.5</u>	<u>-14.2</u>	<u>-12.0</u>	<u>-10.6</u>	<u>-1.9</u>
	<u>South</u>	<u>Short</u>	<u>-2.8</u>	<u>-2.4</u>	<u>-2.0</u>	<u>-1.8</u>	<u>-0.7</u>
		<u>Med</u>	<u>-3.0</u>	<u>-2.4</u>	<u>-2.1</u>	<u>-1.8</u>	<u>-0.8</u>
		<u>Long</u>	<u>-5.2</u>	<u>-2.7</u>	<u>-2.1</u>	<u>-1.9</u>	<u>-0.9</u>
<u>Hardwoods</u>	<u>North</u>	<u>Short</u>	<u>-8.8</u>	<u>-7.1</u>	<u>-6.6</u>	<u>-6.0</u>	<u>-1.4</u>
		<u>Med</u>	<u>-12.2</u>	<u>-7.9</u>	<u>-7.2</u>	<u>-6.5</u>	<u>-3.0</u>
		<u>Long</u>	<u>-24.6</u>	<u>-12.8</u>	<u>-10.0</u>	<u>-8.0</u>	<u>-6.3</u>
	<u>South</u>	<u>Short</u>	<u>-7.1</u>	<u>-3.3</u>	<u>-2.8</u>	<u>-2.4</u>	<u>-1.0</u>
		<u>Med</u>	<u>-8.7</u>	<u>-4.1</u>	<u>-3.1</u>	<u>-2.6</u>	<u>-1.6</u>
		<u>Long</u>	<u>-10.7</u>	<u>-7.2</u>	<u>-5.8</u>	<u>-3.2</u>	<u>-12.0</u>

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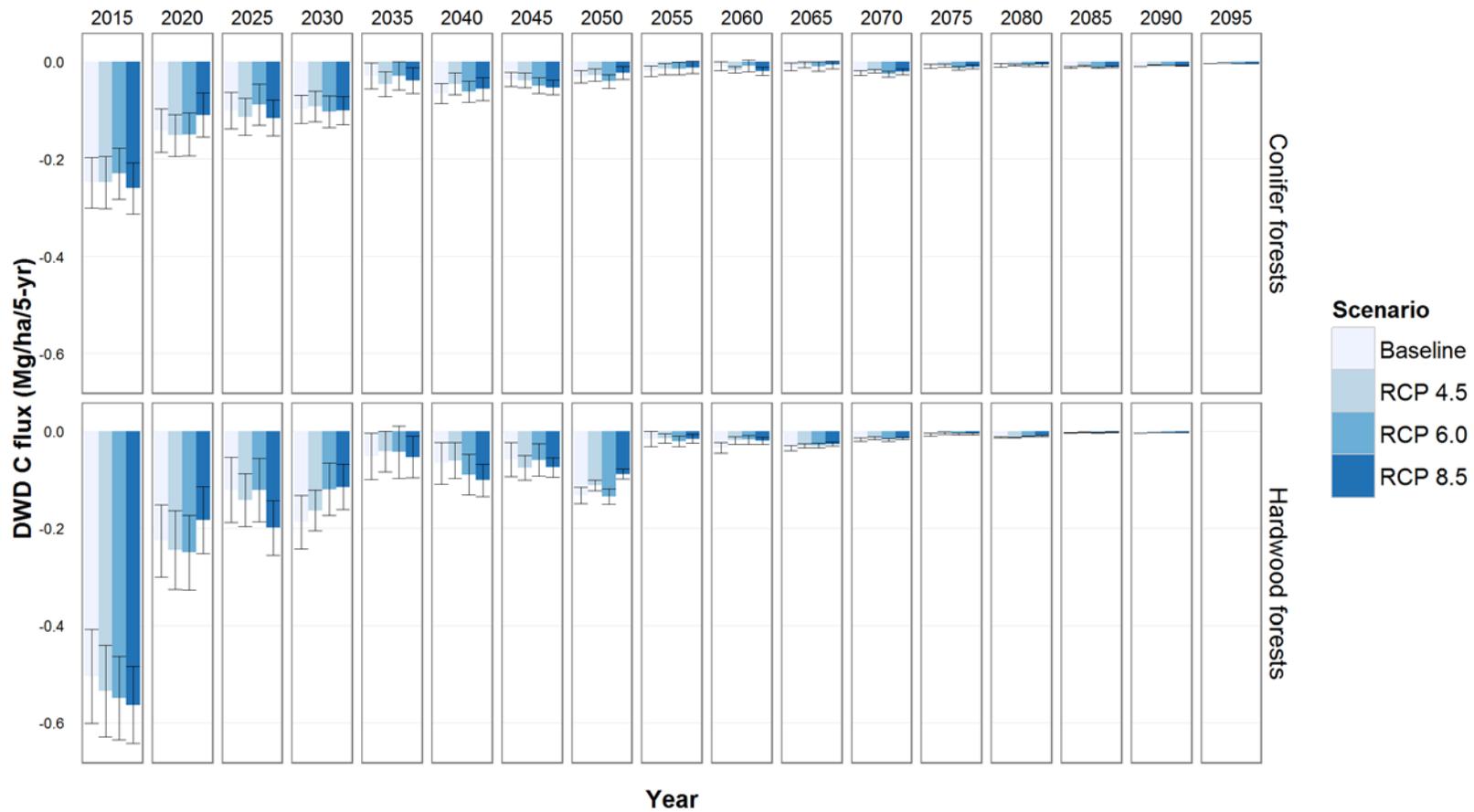


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 2 Figure 1. Violin plots of precipitation and degree day trends for study plots by geographic
 3 region across the eastern US for the climate normal period (1961-1990) and projected
 4 climates using an ensemble of 17 GCMs for CMIP5 models and an RCP 6.0 scenario.



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2 Figure 2. Decreases in downed woody debris (DWD) residence time compared to baseline
3 scenario. Mean values by species group and region in eastern US forests for a baseline current
4 climate scenario and assuming changes in future climate (based on three RCP scenarios from
5 an ensemble of 17 CMIP5 models). Error bars indicate one standard deviation.

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2 Figure 3. Projected downed woody debris carbon flux initialized using the most recent inventory (2007-2011) in eastern US forests for a baseline
 3 current climate scenario and assuming changes in future climate (based on three RCP scenarios from an ensemble of 17 CMIP5 models and not
 4 accounting for future DWD inputs). Error bars indicate \pm one standard error. Conifer forests include loblolly/shortleaf pine, longleaf/slash pine,
 5 spruce/fir, white/red/jack pine, and other softwood forest type groups. Hardwood forests include aspen/birch, elm/ash/cottonwood, maple/beech/birch,
 6 oak/hickory, and other hardwood forest type groups.