Projected gains and losses of wildlife habitat from bioenergy-induced landscape change

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Abstract
Domestic and foreign renewable energy targets and financial incentives have increased demand for woody biomass and bioenergy in the southeastern United States. This demand is expected to be met through purpose-grown agricultural bioenergy crops, short-rotation tree plantations, thinning and harvest of planted and natural forests, and forest harvest residues. With results from a forest economics model, spatially explicit state-and-transition simulation models, and species–habitat models, we projected change in habitat amount for 16 wildlife species caused by meeting a renewable fuel target and expected demand for wood pellets in North Carolina, USA. We projected changes over 40 years under a baseline ‘business-as-usual’ scenario without bioenergy production and five scenarios with unique feedstock portfolios. Bioenergy demand had potential to influence trends in habitat availability for some species in our study area. We found variation in impacts among species, and no scenario was the ‘best’ or ‘worst’ across all species. Our models projected that shrub-associated species would gain habitat under some scenarios because of increases in the amount of regenerating forests on the landscape, while species restricted to mature forests would lose habitat. Some forest species could also lose habitat from the conversion of forests on marginal soils to purpose-grown feedstocks. The conversion of agricultural lands on marginal soils to purpose-grown feedstocks increased habitat losses for one species with strong associations with pasture, which is being lost to urbanization in our study region. Our results indicate that landscape-scale impacts on wildlife habitat will vary among species and depend upon the bioenergy feedstock portfolio. Therefore, decisions about bioenergy and wildlife will likely involve trade-offs among wildlife species, and the choice of focal species is likely to affect the results of landscape-scale assessments. We offer general principals to consider when crafting lists of focal species for bioenergy impact assessments at the landscape scale.

Keywords: biodiversity, bioenergy target, biofuel, habitat, landscape change modeling, renewable energy, southeastern United States, wildlife, wood pellets

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Introduction
In the United States and Europe, government policies encourage the use of bioenergy as an alternative to fossil fuels. In the United States, the Energy Policy Act of 2005 and the Energy Independence and Security Act of 2007 set nationwide targets for a renewable fuel standard whose goals include replacing some dependence on fossil fuels through bioenergy production and reducing greenhouse gas (GHG) emissions (Energy Policy Act, 2005; Energy Independence and Security Act, 2007). The European Union set renewable energy targets that are expected to be met, in part, by harvesting forests in the southeastern United States to produce wood pellets (Abt et al., 2014; Wang et al., 2015; Galik & Abt, 2015). Besides alleviating fossil fuel usage and GHG emissions, bioenergy production can benefit rural, agricultural economies by diversifying markets, increasing demand, and supplementing profits for traditional crops (Dale et al., 2010) or so-called purpose-grown bioenergy crops, such as switchgrass, sorghum, and short-rotation woody crops (SRWCs) composed of pine (Pinus spp.) or poplar (Populus spp.) (Coleman & Stan- turf, 2006).

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As renewable fuel targets and incentives increase bioenergy production, landscapes and the wildlife that inhabit them will be affected through the intensification of agriculture and forest management, influences on land use change (direct and indirect), including conversion of land to forest and agriculture, and the prevention of urbanization (Groom et al., 2008; Robertson et al., 2008; Eggers et al., 2009; McDonald et al., 2009; Dauber et al., 2010; Fargione, 2010; Wang et al., 2015). However, the specific effects of this increase will likely vary across spatial and temporal scales, geographic regions, the wildlife species considered, feedstock types, feedstock management practices, and land use histories (Fletcher et al., 2010; Wiens et al., 2011; Rupp et al., 2012; Immerzeel et al., 2014). The conversion of land from natural ecosystems (e.g., forests, woodlands, grasslands) to ones dominated by crop monocultures could be one of the most important ecological consequences of the increase in demand for bioenergy because such changes are associated with losses of wildlife habitat and biodiversity (Tilman et al., 2002; Firbank, 2008; Dornburg et al., 2010; Fletcher et al., 2010; Wiens et al., 2011; Rupp et al., 2012; Immerzeel et al., 2014). Although some crops grown for bioenergy such as Miscanthus, sweet sorghum, switchgrass, and short-rotation woody crops (SRWCs) provide food and/or cover for some species, the crops differ in their value for wildlife and if grown in monocultures will likely have little value to some species because of their lack of heterogeneity and structure (Fletcher et al., 2010; Wiens et al., 2011; Bonin & Lal, 2012). In addition, crop management practices, such as the timing of harvest relative to animals’ annual and breeding cycles, could also limit their value for wildlife, as well as reduce survival and reproduction (Rupp et al., 2012).

The degree to which land conversions to bioenergy crops negatively affect wildlife depends in part upon the initial land use or land cover. Whereas the conversion of many natural ecosystems to bioenergy crops would represent a loss of habitat value, there may be cases where it improves habitat (Meehan et al., 2010; Wiens et al., 2011; Rupp et al., 2012). For example, if existing monocultures were replaced with structurally heterogeneous grasslands composed of diverse perennial plants, they could support more diverse wildlife communities and increase habitat availability for some grassland species (Rupp et al., 2012; Uden et al., 2015). Such benefits to wildlife would likely be lower in fields with less heterogeneity in plant species composition and structure and lower than in native grasslands, depending upon crop harvest and management practices (Hartman et al., 2011; Wiens et al., 2011; Robertson et al., 2012). SRWC monocultures are associated with lower biodiversity than natural forest ecosystems but may have higher diversity than agricultural croplands (Rupp et al., 2012). Little is known about herpetofaunal diversity on SRWCs (Rupp et al., 2012), but bird and mammal species richness and abundance in midwestern SRWC Populus plantations were lower than in forests and shrublands and higher than in nonhay croplands (Christian et al., 1998).

In the southeastern United States, biomass could also be derived from forests through elevated rates of harvest and thinning and removal of forest harvest residues. The specific characteristics of these harvests will determine their impact on species, although an experimental study failed to detect relationships between the amount of retained woody biomass in clear-cut pine plantations and herpetofauna diversity, evenness, or richness (Fritts et al., 2016) or shrew abundance (Fritts et al., 2015). Decisions regarding which forest types to extract biomass from (e.g., forested uplands vs. wetlands, natural vs. planted stands) will influence the magnitude of effects on individual species (Evans et al., 2013). Thinning forests can increase wildlife diversity by increasing structural complexity, but specific impacts vary across species, thinning technique, and thinning intensity (Rupp et al., 2012).

Land conversions, bioenergy crop cultivation, and forest biomass harvests at individual sites may scale up to biodiversity impacts that are measurable at the landscape level. Once again, the degree and nature of such changes depend upon the initial landscape and bioenergy sources utilized. Heterogeneous landscapes support greater biodiversity (Wiens et al., 2011), so the conversion of abandoned and marginal lands to crop agriculture, along with the concentration of biomass crops around processing facilities, could decrease heterogeneity on some landscapes and, therefore, negatively influence biodiversity (Fletcher et al., 2010; Wiens et al., 2011). Alternatively, planting perennial grasslands for bioenergy feedstocks could be an opportunity to increase the heterogeneity of landscapes that are currently dominated by agricultural crops (Wiens et al., 2011). Effects of the establishment of SRWCs on landscape- and regional-scale biodiversity will vary depending on the degree to which landscapes are already forested and the configuration of SRWC stands, but even relatively small amounts of SRWC plantations in forested landscapes with little existing shrub-dominated land would enhance landscape-scale biodiversity. This effect would not necessarily be the same for landscapes dominated by grasslands or managed forests (Rupp et al., 2012). Rupp et al. (2012) argued that forest thinning for biomass will likely increase the forest structural diversity in landscapes and consequently increase landscape-level species diversity and abundance.

The influence of biomass cultivation and harvest on wildlife habitat at large spatial and temporal scales has
rarely been demonstrated (Immerzeel et al., 2014). Few long-term or landscape-scale studies of bioenergy, effects on wildlife in the southeastern United States exist other than Evans et al. (2013), which examined the concurrence of wildlife habitat and forest biomass harvest risk under scenarios that differed in factors such as the types of forests harvested, the spatially explicit probability of pine plantation establishment, and the intensity of biomass extraction at harvest sites. In our study, we examined the potential impacts of meeting realistic bioenergy demands on habitat availability across a landscape over a 40-year period. While previous large-scale studies examined single sources of biomass for bioenergy, we assessed impacts of bioenergy production under different portfolios of biomass sources that included land conversions to bioenergy crops and harvests of forest biomass. We accomplished this with deductive habitat modeling and results from a timber supply model that was integrated with a spatially explicit state-and-transition simulation model (Costanza et al., 2016). Our results account for complex economic processes that link demand for forest biomass to landscape change, as well as other landscape change drivers, such as urbanization and vegetation dynamics. Furthermore, comparing across different scenarios allowed us to isolate some relationships between specific biomass sources or land conversions and individual species to gain insights into patterns that may emerge when bioenergy demand is applied to landscapes. Our objectives were to (i) determine whether realistic levels of bioenergy demand are large enough to translate into gains or losses in the amount of habitat for individual species at the landscape scale, (ii) assess the degree of similarity in habitat effects among diverse bioenergy portfolios, and (iii) find insightful patterns and connections between specific sources of biomass and individual species. Exploring these topics will inform future efforts to understand how bioenergy demand will impact wildlife and help policymakers craft bioenergy targets and feedstock portfolios that support sustainable wildlife populations.

Materials and methods

Study area

North Carolina (NC) is located in the southeastern United States and includes three distinct physiographic regions, each encompassing different types of land use and vegetation: the Mountains, Piedmont, and Coastal Plain. The majority of publicly owned forested lands lie in the Mountains, which are heavily forested with hardwood and mixed pine and hardwood forests. Most of the urban areas occur within the Piedmont where pine, hardwood, and mixed pine–hardwood forests form a matrix with agricultural lands. Much privately owned agricultural use is spread over the eastern third of the state in the Coastal Plain where diverse plant communities such as bottomland hardwood forests, pine woodlands, and freshwater wetlands also occur. Landscape change due to attempts to meet future timber and agricultural demands, including demands for bioenergy, will likely occur within the Coastal Plain and Piedmont regions. Traditional silviculture that is dominated by loblolly pine (Pinus taeda) plantation growth and management is also common in the Coastal Plain, and much of the future urbanization is predicted to occur in the Piedmont (Terando et al., 2014; Costanza et al., 2015). A state goal of replacing 10% of petroleum sold in NC with locally grown and produced liquid biofuels through 2017 was chosen in 2007 (Burke et al., 2007). Additionally, existing or announced wood pellet plants could consume 3.45 million green tonnes (3.8 million green short tons) of forest biomass annually (Forisk Consulting LLC, 2014). We used deductive species–habitat modeling to translate future projected landscapes into potential trends in habitat availability within NC between 2010 and 2050 for 16 wildlife species under five alternative bioenergy production scenarios.

Bioenergy scenarios and land use/land cover

In the southeastern United States, bioenergy will likely be produced from agricultural crop residues, bioenergy crops (e.g., sorghum and switchgrass), SRWC, the harvest or thinning of forests, and forest harvest residues (Rupp et al., 2012; Fritts et al., 2014; Wang et al., 2015). We could not predict the exact portfolio of sources in NC’s future, so we explored Costanza et al.’s (2016) five bioenergy production scenarios along with their baseline that did not include increased bioenergy production (Table 1). The bioenergy scenarios differed in how much they utilized three sources of biomass: (i) feedstock sourced from the harvest and thinning of planted pine, natural pine, hardwood, mixed pine–hardwood, upland, and bottomland forests (hereafter, ‘conventional forestry’); (ii) the conversion of agricultural lands on poor soils to purpose-grown feedstocks and subsequent harvests (i.e., switchgrass, sorghum, SRWC; hereafter ‘conversion of Ag lands’); and (iii) the conversion of forests on poor soils to purpose-grown feedstocks and subsequent harvests (hereafter, ‘conversion of forests’).

One of the bioenergy scenarios (‘Conventional’) assumed that 3.63 million green tonnes (4.0 million green short tons) of forest biomass would be harvested, and thus met the anticipated demand for biomass for wood pellet production (Table 1). Two other scenarios (‘Conventional-Ag’, and ‘Conventional-Ag-Forest’) also included the harvest of 3.63 million green tonnes of forest biomass for bioenergy, as well as enough purpose-grown feedstocks to meet or exceed the 10% liquid if the forest biomass was used for biofuels instead of wood pellets (Costanza et al., 2016). Those two scenarios differed in the types of lands converted to purpose-grown feedstocks. Conventional-Ag assumed that only marginal agricultural lands would be converted to purpose-grown crops, whereas Conventional-Ag-Forest assumed that marginal agricultural and forest lands, in equal proportions, would be converted to purpose-grown crops. The remaining two bioenergy scenarios (‘Ag’ and
amounts that would result in 3.63 million green tonnes of
mass, the SRTS model projected forest harvest and thinning
For the three scenarios that included conventional forest bio-
USDA Forest Service Forest Inventory and Analysis (FIA) data.
Including Baseline), the SRTS model incorporated demand for
conversion between agricultural and forest land uses. For the
ies of areal amounts of forest thinning and harvest, as well as
and-transition simulation model (STSM) software ST-Sim
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ies of areal amounts of forest thinning and harvest, as well as
conversion between agricultural and forest land uses. For the
three scenarios without biomass from conventional forests (in-
cluding Baseline), the SRTS model incorporated demand for
broad forest types and age classes according to recent trends in
USDA Forest Service Forest Inventory and Analysis (FIA) data.
For the three scenarios that included conventional forest bio-
mass, the SRTS model projected forest harvest and thinning
amounts that would result in 3.63 million green tonnes of

| Scenario name | Description | Conv. forest* (%) | Marg. Ag.† (%) | Marg. forest‡ (%) | NCRFS§ (%) | Forest biomass¶ (M gt) | Harvest residue** (%) | Ag. converted†† (ha) | Forest converted‡‡ (ha) |
|---------------|-------------|------------------|---------------|-----------------|------------|---------------------|---------------------|-------------------|---------------------|
| Baseline      | Business-as-usual production of conventional forest products; no bioenergy production | 0               | 0             | 0               | 0          | n/a                 | 0                   | 0                 | 0                   |
| Conventional  | Increased conventional forestry for bioenergy | 100             | 0             | 0               | 0          | 3.63                | 40                  | 0                 | 0                   |
| Conventional-Ag | Increased conventional forestry and conversion of agricultural lands for bioenergy | 15              | 85            | 0               | 100–148    | 3.63                | 40                  | 425 000            | 0                   |
| Conventional-Ag-Forest | Increased conventional forestry and conversion of agricultural lands and forests for bioenergy | 15              | 42.5          | 42.5            | 100–148    | 3.63                | 40                  | 212 500            | 212 500             |
| Ag            | Conversion of agricultural lands for bioenergy | 0               | 100           | 0               | 100–140    | 0                   | n/a                 | 504 000            | 0                   |
| Ag-Forest     | Conversion of agricultural lands and forests for bioenergy | 0               | 50            | 50              | 100–140    | 0                   | n/a                 | 252 000            | 252 000             |

*Percentage of scenario derived from planted and natural forests.
†Percentage of scenario derived from conversion of agricultural lands on marginal soils to purpose-grown feedstocks.
‡Percentage of scenario derived from conversion of forests on marginal soils to purpose-grown feedstocks.
§Potential percentage of North Carolina’s 10% biofuel target met (see text for description of 10% target). Ranges are presented instead of point estimates because the amount of energy that can be derived from feedstocks varies with refinement process (Costanza et al., 2016).
¶Biomass from conventional forestry (in million green tonnes).
**Percentage of forest harvest residues removed.
††Amount of agricultural land converted to switchgrass, sorghum, or SRWC.
‡‡Amount of forest land converted to switchgrass, sorghum, or SRWC.

‘Ag-Forest’) would meet the 10% biofuel target without increased conventional forestry. Ag assumed that marginal agricultural lands would be the only lands converted to purpose-grown crops, whereas Ag-Forest incorporated conversion of equal proportions of marginal agricultural and forest lands. All five bioenergy scenarios and the baseline scenario included urbanization, natural disturbance such as fire, ecological succession, and conversion between forest and agricultural land use according to recent change trends (Wear & Greis, 2013; Costanza et al., 2016).

Costanza et al. (2015, 2016) translated demand for bioenergy into potential changes in NC’s land cover using the Sub-regional Timber Supply Model (SRTS) (Abt et al., 2009) and state-and-transition simulation model (STSM) software ST-Sim (ApexRMS, 2014). The SRTS model produced aspatial time series of areal amounts of forest thinning and harvest, as well as conversion between agricultural and forest land uses. For the three scenarios without biomass from conventional forests (including Baseline), the SRTS model incorporated demand for broad forest types and age classes according to recent trends in USDA Forest Service Forest Inventory and Analysis (FIA) data. For the three scenarios that included conventional forest biomass, the SRTS model projected forest harvest and thinning amounts that would result in 3.63 million green tonnes of biomass extracted from forests, while accounting for demand displacement of other forest products (Costanza et al., 2015).

The spatial STSM incorporated time series outputs of forest thinning, harvest, and land use conversion from SRTS and aspatial state-and-transition model pathways for each of 52 vegetation and land use classes. It also included several spatial inputs: maps of initial vegetation, land use, successional stages, and vegetation structure; an existing time series of future urbanization probability (Terando et al., 2014); marginal agricultural and forest lands to be converted to purpose-grown crops for bioenergy; and lands under conservation ownership to be excluded from biomass production (Costanza et al., 2016). To ensure that adjacent pixels were more likely to undergo disturbance, management, and land conversion transitions together, rules were set in the STSM that governed the sizes of these events so that no transitions were smaller than 2.0 ha (approximately six pixels), which was approximately the minimum patch size in the initial vegetation and land use raster for NC (Costanza et al., 2016).

Initial vegetation and land use maps and most of the aspatial model pathways are detailed in Costanza et al. (2015). Costanza et al. (2016) added model pathways corresponding to each of the three purpose-grown crops. Switchgrass and sorghum pathways each consisted of a single state class, whereas the
SRWC pathway contained three state classes: a recently harvested class, an early-succession class, and a mid-succession class. When lands were converted to SRWC, they transitioned to the early-succession class. The planted pine pathway was divided into nine state classes: one recently harvested class, three recently thinned classes, and five classes that represented various age classes of closed (not recently thinned) forest (Table S1). The accuracy of the initial vegetation and land use map has not been assessed, but some accuracy metrics have been measured for the datasets from which it was created (Costanza et al., 2016). The current version of the GAP Land Cover Map has not been formally assessed, but a similar land cover mapping effort for the North Carolina Gap Analysis had per-class accuracies between 74% and 95% for general land cover classes and overall accuracy of 58% for the detailed classes based on conditional probabilities (McKerrow et al., 2006). Validation of the NLCD canopy cover model with canopy cover values from field data by Coulston et al. (2012) showed an $R^2$ value of 0.8 in the southeastern United States. LANDFIRE s-class data are based on LANDFIRE vegetation height data, which have a spatial bias of 3.8%, and LANDFIRE cover data had an overall agreement of 74% when compared with field data across the U.S. (Toney et al., 2012).

Costanza et al. (2016) identified marginal agricultural and forest lands as those that were row crop, pasture/hay, or forest in NLCD 2006, and were in nonirrigated soil capability classes 3 or 4 according to gridded Soil Survey Geographic Database (gSSURGO). They used GAP PAD-US data (U.S. Geological Survey National Gap Analysis Program, 2013) to identify lands under conservation ownership that would be ineligible for biomass production and conversion to urban or agricultural land. The STSM simulated changes in each of the mapped vegetation and land use classes due to forest management and land conversions specific to bioenergy production described above, as well as natural disturbance processes, ecological succession, urbanization, forest management, and background conversion between forest and agricultural lands.

The spatial outputs resulting from the STSMs had 60 m × 60 m cells. Each cell was assigned a land use or vegetation class. Vegetation classes were also assigned a successional stage (early, mid, or late succession) and a structure class (open or closed canopy, thinned, or recently harvested) (Table S2). Hereafter, we refer to these unique combinations of vegetation class, successional stage, and structure class as ‘land cover classes’. Whereas SRTS and the STSMs were run on an annual time step, spatial outputs were produced every 10 time steps. In order to match the resolution of other model inputs, we resampled these spatial vegetation and land use outputs to 30 m × 30 m cell resolution using a nearest neighbor technique that assigned the values of larger cells to the each of the smaller, nested cells and used these finer resolution raster layers in the habitat models.

**Focal species**

While modeling all terrestrial vertebrate species within NC would have been ideal for our study, doing so was impractical given the effort and information required to build a habitat model. Therefore, we carefully selected 16 species that, collectively, allowed us to meet our objectives, focusing on birds and amphibians because numerous published studies on the habitat associations of these taxa existed (Table 2). To meet our objective of determining whether bioenergy demand could translate into changes in habitat amount, given the economic and ecological processes at play on the landscape, we sought a list that included species that were, collectively, associated with each of the major systems that will be affected by bioenergy demand (e.g., forests, woodlands, and agricultural lands) and could be sensitive to the changes in successional stage and vegetative structure likely resulting from biomass harvests, such as decreases in percent canopy cover or the development of dense shrub layers. To meet our objective of exploring the diversity of responses, we sought a list of species with diverse habitat preferences. Therefore, we chose species that contributed a unique suite of land cover class associations with the list and that select habitat in part on the basis of vegetative composition or structure at the level of detail concomitant with the land cover classification and changes in forest successional stage and vegetative structure in the STSM. We presumed species which do not select habitat based on vegetative structure would be unlikely to experience changes in habitat amount from forest biomass harvests and did not consider them for inclusion. Similarly, landscape change would be less likely to manifest in changes in habitat availability for habitat generalists than for habitat specialists, so we avoided adding individual species associated with the majority of the land cover classes in the land cover map. Our list included 10 species of greatest conservation need, three species of special concern, and two species listed as threatened by NC (NCWRC, 2015; Table S4). In addition, seven are on the Partners in Flight Watch List and four are on the IUCN Red List (Rich et al., 2004; IUCN, 2015; Table S4).

**Habitat models**

We built a deductive habitat model for each species that categorized STSM land cover classes as either inhabitable or uninhabitable in order to estimate future trends in species’ habitat availability under each scenario (Table S2). Our deductive approach (habitat models) followed that of the United States Geological Survey’s National Gap Analysis Program (GAP; Scott et al., 1993; Corsi et al., 2000; Overmars et al., 2007). GAP models are based on species–habitat requirement information gleaned from scientific literature and expert opinion. They associate individual species with land cover/land use classes (Gergely & McKerrow, 2013) and incorporate ancillary parameters (e.g., elevation, distance from forest edge, and distance from water) and range data to map the distribution of individual species’ habitat at a landscape scale (1:100 000 map scale). Whereas national GAP models are based upon the GAP National Land Cover, ours corresponded with the land cover classification used in STSM, which has a finer thematic resolution than GAP’s classification due to the specification of multiple states and stages (Table S2). For several species, we added constraints on model output related to maximum elevations and distances to water features as specified by the species’ GAP model (Table S5) (http://gapanalysis.usgs.gov/). We
We built deductive habitat models for 16 focal species to explore bioenergy demand effects on wildlife habitat in North Carolina. X's indicate that the species is associated with a land cover or vegetative structure type, but we caution that the associations listed here are highly generalized from the associations in the habitat models, which are described in Tables S2 and S3 and include specification of suitable successional stages and vegetative structure classes. Percent habitat refers to the proportion of the regions listed here are highly generalized from the associations in the habitat models, which are described in Tables S2 and S3 and include specification of suitable successional stages and vegetative structure classes. Percent habitat refers to the proportion of the land area of North Carolina that our species–habitats models predicted as suitable for the species in 2010.

### Model assumptions

Our modeling methods necessarily involved a set of assumptions about bioenergy feedstocks and habitat use patterns because some feedstocks are not yet widely present in NC or their characteristics and suitability for wildlife have not been documented. First, we assumed that forest thinning, whether under Baseline or in the bioenergy scenarios, removed 50% of biomass and therefore 50% of canopy cover. In the SRTS model, forest thinning was largely used for nonbioenergy products including sawtimber, so larger diameter trees would be included in the thinnings with and without bioenergy demand. Second, we assumed forest harvest under bioenergy scenarios removed 40% of residual biomass while retaining some snags (Costanza et al., 2015). Third, Conventional-Ag, Conventional-Ag-Forest, Ag, and Ag-Forest included the cultivation of sorghum, switchgrass, and SRWC, but the extent to which wildlife will use these crops in NC has not been thoroughly studied. We assumed that none of our species will breed in sweet sorghum because they rarely breed in traditional row crops in NC; switchgrass crops will be tall, dense monocultures that may only be suitable for Yellow-breasted Chats; and our focal species would inhabit SRWC at rates similar to those in young planted pine stands (Dhondt et al., 2007).

### Analysis and summary of habitat model outputs

The six scenarios (Baseline and five bioenergy portfolios) and five time steps (10-year increments from 2010 to 2050) yielded...
30 separate model outputs for each of 16 focal species. We quantified the influence of each bioenergy feedstock portfolio on habitat availability for each species by calculating the differences in the amounts of habitat available between each bioenergy scenario and Baseline. To do this, we converted the number of cells in the habitat model output from each species-scenario-time step combination into hectares (Table S6) and then calculated the areal differences between individual bioenergy scenarios and baseline values at each time step for each species. We also summarized changes in habitat amount under Baseline in order to provide background references for both the amount of habitat initially available and projected trends given drivers of landscape change other than bioenergy production. However, we limit our presentation of baseline trends because they were somewhat difficult to interpret for some species due to modeled increases in new thinned and harvested land cover classes on the landscape during the first ten annual time steps of the STSM (Costanza et al., 2015). Those large modeled landscape changes had the capacity to affect trends for species that use intact, thinned, or harvested forests during the same time period. We discuss this issue further in the Discussion.

**Results**

Our habitat models estimated that in 2010, the proportion of NC that was suitable ranged from <1% to 54% among our focal species (Table 2). For most species, between 11% and 20% of the state was suitable, but only three to four percent was suitable for the Prothonotary Warbler and Eastern Tiger Salamander, one percent for the Oak Toad and Mole Salamander, and less than one percent for the Gopher Frog. Among the species with the most habitat were Kentucky Warbler (27%), Brown-headed Nuthatch (33%), Wood Thrush (41%), and Hairy Woodpecker (54%).

Under Baseline, trends in the projected amount of habitat varied by species. Habitat was lost with each time step for the Eastern Tiger Salamander, Mole Salamander, Cerulean Warbler, Grasshopper Sparrow, Hairy Woodpecker, Swainson’s Warbler, and Wood Thrush (Fig. 1). In addition, a net loss of habitat occurred by 2050 for the Kentucky Warbler, Loggerhead Shrike, and Prothonotary Warbler, but these species experienced gains and losses during the 40-year period. The Gopher Frog also experienced gains and losses during the 40-year period, but 2050 habitat amounts were similar to initial amounts. Mole and Eastern Tiger Salamanders experienced habitat losses in both upland and bottomland hardwood and conifer forests. Habitat was gained by 2050 for the Red-headed Woodpecker, Brown-headed Nuthatch, Oak Toad, Yellow-breasted Chat, and Field Sparrow (Fig. 1).

For several species, trends in habitat amount remained similar to Baseline under scenarios with bioenergy production; that is, the amount of habitat available did not deviate more than five percent from the amount available under Baseline at any time step under any scenario (Fig. 2). Nevertheless, some patterns emerged for those species. Conventional, Conventional-Ag, and Ag were similar to Baseline for the Mole Salamander, Brown-headed Nuthatch, Eastern Tiger Salamander, Hairy Woodpecker, and Wood Thrush, but Conventional-Ag-Forest and Ag-Forest, which included the conversion of marginal forests, offered less habitat than Baseline for those species. Cerulean Warbler habitat amounts were most similar to Baseline under scenarios without increased conventional forestry (Ag and Ag-Forest) and declined over the 40-year period in scenarios with increased conventional forestry. However, those declines were never more than 4% below Baseline. Kentucky Warbler habitat amounts were most similar to Baseline under the scenario without conversion of agricultural lands or forests (Conventional) but alternated gains and losses of up to 5% from Baseline in scenarios with conversion of agricultural lands or marginal forests. Habitat for Swainson’s Warbler also varied among time steps under each bioenergy scenario, but never deviated more than 4% from Baseline.

Among species that experienced >5% deviation from Baseline under a bioenergy scenario, the trends and magnitudes of differences varied among species, scenarios, and time steps (Fig. 2). The Gopher Frog, Grasshopper Sparrow, Loggerhead Shrike, and Yellow-breasted Chat experienced the greatest differences from Baseline with Gopher Frog habitat falling 20–25% below Baseline by 2050 under scenarios that included increases in conventional forestry. Grasshopper Sparrow habitat area was almost 20% below Baseline and Loggerhead Shrike habitat area was 12% below Baseline under Ag, in which all bioenergy was derived from the conversion of agricultural lands to purpose-grown feedstocks. The Yellow-breasted Chat gained 14% more habitat by 2040 under scenarios that included added conventional forestry and conversion of agricultural lands to bioenergy feedstocks.

Several noteworthy relationships between specific bioenergy feedstocks and species’ habitat were identifiable. First, for a few species, the trends in habitat availability under scenarios with increased conventional forestry differed from trends under the other scenarios. For the Oak Toad, those scenarios were better than all other scenarios before 2040, although in 2050, scenarios without added conventional forestry (Ag and Ag-Forest) were also better than Baseline (4% and <1%, respectively). For the Gopher Frog, which has similar habitat associations with the Oak Toad except that it is not associated with hardwood forests (Table 2), increasing conventional forestry decreased the habitat amount to as much as 25% below Baseline in 2050 (25% under...
Conventional-Ag-Forest; 23% under Conventional; 20% under Conventional-Ag). For the Prothonotary Warbler and the Red-headed Woodpecker, scenarios that included increased conventional forestry caused greater deviation from Baseline while remaining similar to each other, whereas habitat amounts under the other scenarios (Ag and Ag-Forest) were very similar to Baseline. However, Prothonotary Warbler habitat decreased to almost 10% below Baseline, whereas Red-headed Woodpecker habitat increased to around 5% above Baseline by 2030 and then decreased to 2–3% above Baseline by 2050 under Conventional, Conventional-Ag, and Conventional-Ag-Forest. Scenarios that included increased conventional forestry were better for the Field Sparrow than scenarios that did not, and converting marginal forests to bioenergy feedstocks also increased the amount of its habitat. Conventional provided a steady gain in habitat relative to Baseline, but habitat amounts under the other scenarios varied among time steps; Conventional-Ag-Forest was consistently better than Baseline with up to 8% more habitat, whereas Ag was consistently worse with 8% less habitat in 2030.

Second, two species clearly responded negatively to the conversion of Ag lands to purpose-grown feedstocks: the Grasshopper Sparrow and the Loggerhead Shrike. For the Grasshopper Sparrow, sourcing biomass for wood pellets (Conventional) did not cause added habitat loss, but the conversion of agricultural lands did; ranking scenarios based on the proportion of bioenergy derived from agricultural land conversion (i.e.,
Conventional, Conventional-Ag-Forest, Ag-Forest, Conventional-Ag, Ag; Table 1) corresponded to their ranking in terms of habitat loss (Fig. 1; Conventional-Ag-Forest and Ag-Forest between 5% and 10%; Ag and Conventional-Ag between 13% and 19%). Patterns in the differences among scenarios for the Loggerhead Shrike were similar to the Grasshopper Sparrow, but muted; the worst scenario, Ag, was 12% below Baseline by 2050.

Finally, all bioenergy scenarios provided more habitat for the Yellow-breasted Chat than Baseline. This species experienced a steady increase relative to Baseline over the 40-year period with increased conventional forestry to meet wood pellet demand and was...

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4% greater than the Baseline in 2050. Under scenarios capable of meeting the biofuel target (Conventional-Ag, Conventional-Ag-Forest, Ag, and Ag-Forest), habitat amounts were between 7% and 13% greater than Baseline in 2050.

Discussion

We projected future changes in habitat availability for 16 species under alternative scenarios of bioenergy production and found that realistic levels of demand for biofuels and wood pellets in NC would be capable of causing considerable changes in the amount of habitat available for some species. These changes could be positive or negative, depending upon a species’ habitat associations and the sources of biomass utilized to meet demand. Several patterns emerged that will inform future research and policy regarding bioenergy and wildlife.

Species that inhabit dense, shrubby vegetation gained habitat in some bioenergy scenarios because forest biomass harvests (clear-cutting and thinning) encouraged development of shrubby vegetative structure (shrubs and saplings), and conversions of forests on marginal land to purpose-grown crops converted unsuitable forests to dense bioenergy crops that we assumed would be suitable for shrubland species. For example, the Field Sparrow experienced habitat gains when conventional forestry was increased, and the Yellow-breasted Chat gained habitat by 2050 under each of the bioenergy scenarios. Chats are associated with open, shrubby areas characteristic of regenerating and thinned forests (Eckle & Thompson, 2001), and they were associated with SRWC and switchgrass in our models. Furthermore, it was unclear whether future chat habitat would be gained or lost compared with 2010 under Baseline because of their sensitivity to modeled landscape changes described below in uncertainty and limitations, but the four scenarios that included the harvest and thinning of these forest types (increased conventional forestry). Similarly, Evans et al. (2013) found that large amounts of forested wetlands are likely to be harvested in NC’s Coastal Plain and among their indicator species, the Prothonotary Warbler had the greatest relative risk of habitat changes due to biomass harvest for wood pellets. There was also a large amount of overlap between Swainson’s Warbler habitat and areas of high harvest risk when forested wetland harvests were allowed in their models, but we found little difference from Baseline for the Swainson’s Warbler under any bioenergy scenario. Our models indicated a response similar to Prothonotary Warbler’s for the Cerulean Warbler, but the magnitude of changes relative to Baseline was smaller. However, this muted response may have been a result of the Cerulean Warbler’s distribution within NC. The species mostly occurs in the Mountains, with a small, isolated population inhabiting forests along the Roanoke River in the Coastal Plain (Fig. S1). Any changes for the coastal population were likely diluted by habitat dynamics occurring in the Mountains. Had our area of interest for this study been limited to the Coastal Plain, results for the Cerulean Warbler may have been different. The case of these bottomland hardwood-associated species suggests a need for further research and monitoring of how species depend upon them and underscores the need to identify important vegetative communities and understand their value to species when crafting policy that could result in changes to the landscapes where they occur.

Converting pasture and croplands on poor soils to purpose-grown bioenergy feedstocks exacerbated habitat loss for two pasture-associated species, the Grasshopper Sparrow and Loggerhead Shrike. These
results support concerns about turning to marginal agricultural lands for bioenergy feedstock cultivation (Meehan et al., 2010; Wiens et al., 2011). Grasshopper Sparrow and Loggerhead Shrike habitats, which are characterized by open, grassy conditions (Vickery, 1996; Yosef, 1996), were lost under all scenarios that included the conversion of marginal agricultural lands to purpose-grown feedstocks. That is, the greater the proportion of bioenergy that came from conversion of marginal agricultural lands, the worse the scenario was for these species. This result was specifically associated with the conversion of pasture located on marginal soils. Our analysis indicates a clear relationship where increasing the cultivation of marginal lands for bioenergy would decrease the amount of habitat for grassland-associated species. Such a decrease is troubling because, in NC, pasture is already expected to be lost with urbanization, a process which was evident in loss of habitat under our ‘business-as-usual’ scenario. Furthermore, grassland birds have undergone the steepest national population decline of any group of birds in the last 40 years (Wiens et al., 2011; Rupp et al., 2012), and species associated with open vegetation may be most vulnerable to future land use change (Martinuzzi et al., 2015). The importance of two drivers of landscape change for the Grasshopper Sparrow in our study indicates that considering bioenergy impacts within the context of other drivers of landscape change, rather than alone, will improve the accuracy of assessments of their impacts on biodiversity.

The conversion of forests located on poor soils to purpose-grown bioenergy feedstocks, including SRWC, was beneficial to the Yellow-breasted Chat, Field Sparrow, and Kentucky Warbler which will presumably inhabit dense, shrubby growth, and/or low canopy cover that will likely exist in SRWC stands. Conversely, converting forests resulted in habitat loss for forest species, such as the Eastern Tiger Salamander, Mole Salamander, Brown-headed Nuthatch, Hairy Woodpecker, Wood Thrush, and Swainson’s Warbler. The diverse responses of species to land conversions in our study echo a point that others have made that impacts of land conversions on wildlife depend upon the initial types of land cover involved in the conversions (Meehan et al., 2010; Wiens et al., 2011; Rupp et al., 2012). Our results suggest that the choice of which species to examine deserve to be added to the list of factors that influence whether land conversions for bioenergy will be deemed ‘good’ or ‘bad’.

By virtue of numbers, species with small ranges, such as rare and range restricted species, are more sensitive to small changes in habitat acreage than common species with large ranges. We estimated that <1% of NC was suitable in 2010 for the Gopher Frog. This species has undergone large range-wide population declines (Jensen & Richter, 2005) and is listed as near threatened by the IUCN and a species of special concern within NC (Hammerson & Jensen, 2004; North Carolina Wildlife Resources Commission, 2015). The overall losses that emerged suggest increased conventional forestry within its range would challenge conservation of this species. However, the initial increase in habitat over Baseline from additional conventional forestry suggests that forest biomass harvests for bioenergy could benefit the species if closed canopy forests were thinned in ways appropriate for it (Wilson, 1995). Little of the state was potentially suitable for the Prothonotary Warbler in 2010 (3%), and whereas its association with mature bottomland hardwood forests is very different than the Gopher Frog’s association with open forests and woodlands on sandy soils, habitat amount for this species also declined (10%) under scenarios with increased conventional forestry. Other species for which habitat was less abundant in 2010 could fare better; the Oak Toad could gain habitat relative to baseline projections by 2050 under all bioenergy scenarios. Mole and Eastern Tiger Salamanders could gain habitat by 2050 under scenarios with increased conventional forestry. These patterns underscore that consideration of the spatial, temporal, and vegetative characteristics (i.e., successional stage and vegetative structure) of landscape changes due to bioenergy demand would benefit conservation assessments of populations of rare species because they have the capacity to emerge as important influences on habitat amounts. Our results suggest that bioenergy policy guidelines should consider the spatial coincidence of biomass production and species that are rare or have small ranges.

Uncertainty and limitations

Some aspects of future bioenergy demand, biomass source characteristics, and species’ habitat use were uncertain in NC, even for the well-described species that we examined. We, therefore, had to make some assumptions in our modeling. The utility of novel bioenergy crops, thinned forests, and harvested forests to wildlife cannot be accurately modeled because their composition and structure will be dictated by cultivation and harvest practices that have not yet been fully developed and described. We, therefore, associated species with some land cover classes under assumptions of those classes’ characteristics. If those assumptions prove inaccurate, then the accuracy of our projections will suffer. For example, if all snags are generally removed from clear-cuts, then Red-headed Woodpeckers will not gain as much habitat from increased conventional forestry as we indicated. Similarly, if switchgrass crops in
NC turn out to have vegetative composition or structure that is unsuitable for the Yellow-breasted Chats, then the species may not gain as much from bioenergy production as we projected. In light of these necessary assumptions, we caution against applying our results to questions other than those that we have addressed, or treating them as predictions. Future work to thoroughly describe and document the characteristics of bioenergy crops and the wildlife species that inhabit them will greatly benefit wildlife impact assessments.

Our study assessed general changes to the landscape, but the importance of finer-scale habitat factors should not be underestimated. For wildlife, the value of a site depends upon site, patch, and landscape characteristics; local management practices; and a myriad of other ‘natural’ influences on individual survival and reproduction not captured in our models. These factors could render patches of suitable land cover only marginally suitable or ecological traps (Schlaepfer et al., 2002). Similarly, creation of habitat via bioenergy crops and forest harvests will depend upon patch characteristics (e.g., field or forest patch size and shape), management practices, and landscape composition (Robertson et al., 2011, 2013; Blank et al., 2014). In forests, influences on survival and reproduction could include predation on adults, juveniles, and nests; nest parasitism; and prey densities (Wilcove, 1985; Streby et al., 2011; Zitske et al., 2011). In crops, management practices and field characteristics are still uncertain (Rupp et al., 2012; Blank et al., 2014; Immerzeel et al., 2014), but management practices including tillage; crop residue harvest intensities; and the intensity, frequency, and timing of harvesting will be important (Rupp et al., 2012). Furthermore, without plant diversity and heterogeneity within fields, bioenergy crops could be of limited value to wildlife (Bonin & Lal, 2012). While we knew that patch and landscape characteristics influence whether sites are used by some of our focal species, we could not include these parameters in our models because predicting the sizes, shapes, and configurations of future habitat patches was not possible in our modeling framework. Understanding site-scale relationships between feedstocks and habitat quality better, as well as increasing the ability to predict future patch characteristics and landscape trends would improve our ability to understand the impacts of bioenergy on habitat (Immerzeel et al., 2014).

Our baseline projections of habitat availability changed more during the first time step than for subsequent time steps for some species (Fig. 1). For example, our model for the Gopher Frog projected large initial gains (120% by 2020) that were lost by 2050, with the amount of habitat in 2050 lower than 2010. We suspect that this was in part an amplification of landscape changes in the first few time steps of the STSM during which new thinned and harvested land cover classes increased on the landscape (Costanza et al., 2015) in combination with the species’ distribution and the aging of recently harvested longleaf pine woodlands in the modeled landscape. These modeled landscape changes had the capacity to affect baseline trends for our species that use intact, thinned, or harvested forests and may have amplified early losses under Baseline for the Mole Salamander, Hairy Woodpecker, Eastern Tiger Salamander, Cerulean Warbler, Prothonotary Warbler, Swainson’s Warbler, and Wood Thrush. There may have been a greater effect on gains and losses by 2050 for other species. For example, habitat was more abundant in 2050 than 2010 under Baseline for the Oak Toad, Gopher Frog, Field Sparrow, and Yellow-breasted Chat, but this overall gain in 2050 may have existed because of the large increase from 2010 to 2020 for those species (Fig. 1).

General insights for research, conservation, and policy

Biomass harvests and land conversions have the capacity to add or remove large amounts of habitat from a landscape under realistic levels of demand for bioenergy. Such changes will occur against a backdrop of other drivers of landscape change, and there may be cases where bioenergy will amplify or mediate changes in habitat amount from other change agents. Future increases in bioenergy production will likely have a variety of implications for wildlife because of the diversity of species’ habitat requirements and biomass sources, along with the importance of how biomass harvests and cultivation will overlap with species’ distributions. Each species that we examined experienced some variation among the bioenergy feedstock portfolios that we explored, but one species benefited under all bioenergy scenarios. None of the scenarios that we explored were ‘best’ or ‘worst’ for all of our focal species, and the impacts of scenarios varied over time for some.

Our models indicated that there will be both ‘winners’ and ‘losers’ under a given bioenergy portfolio, and how a species fares may vary over time. Consequently, bioenergy policies will likely involve trade-offs for wildlife, and choices about which species to include in assessments of wildlife impacts will influence the assessments’ conclusions. When evaluating the sustainability implications of bioenergy production, researchers and policymakers should therefore carefully consider whether to focus on all species, or subgroups of species occurring in an area of interest. If subgroups are examined, then the groups should be carefully defined to match the objectives. We present four general principles to consider when undertaking such a task, although there are likely others. One, species that inhabit regenerating forests may benefit from bioenergy demand. Two, species that

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are largely dependent upon the mature state of a single land cover type that could be harvested for biomass are at risk to be negatively impacted by bioenergy demand. Three, bioenergy demand could exacerbate habitat loss for species that primarily inhabit land cover types that are also being lost to other drivers of landscape change (e.g., urbanization). Four, species with small ranges can be more sensitive to landscape changes from bioenergy and deserve special consideration.

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References

Abt RC, Cubbage FW, Abt KL (2009) Projecting southern timber supply for multiple products by sub region. Forest Products Journal, 59, 7–16.

Abt KL, Abt RC, Galik CS, Skog KE (2014) Effect of policies on pellet production and forests in the U.S. South: a technical document supporting the Forest Service update of the 2010 RFA Assessment. Gen. Tech. Rep. SRS-202, U.S. Department of Agriculture Forest Service, Southern Research Station, Asheville, NC, 33 p.

ApeX Resource Management Solutions (2014) ST-Sim state-and-transition simulation model software. Available at: http://www.apexrms.com/stsim (accessed 1 July 2014).

Blank PJ, Sample DW, Williams CL, Turner MG (2014) Bird communities and biofuel crops: results and modeling in potential bioenergy grasslands. PLoS ONE, 9, 1–10.

Bozin C, Lal R (2012) Agronomic and ecological implications of biofuels. In: Advances in Agronomy, Vol 117 (ed. Donald LS), pp. 151–190, Academic, New York, USA.

Burke S, Hall BR, Shahbazi G, Tolson EN, Wynne JC (2007) North Carolina’s forests and forest landscape change in the southeastern U.S. Simulation and timber supply models for forest biomass production scenarios. 11 December 2015).

Christian DP, Hoffman W, Hanovitch MJ, Niemi GJ, Beyea J (1998) Bird and mammal diversity on woody biomass plantations in North America. Biomass and Bioenergy, 14, 395–402.

Coleman MD, Stanturf JA (2016) Biomass feedstock production systems: economic and environmental benefits. Biomass and Bioenergy, 90, 693–695.

Corso F, de Leeuw J, Skidmore A (2010) Modeling species distribution with GIS. In: Research Techniques in Animal Ecology (eds Bottani L, Fuller T), pp. 389–434, Columbia University Press, New York.

Costanza JK, Abt RC, McKerrow AJ, Collazo JA (2015) Linking state-and-transition simulation and timber supply models for forest biomass production scenarios. AIMS Environmental Science, 2, 180–202.

Costanza JK, Abt RC, McKerrow AJ, Collazo JA (2016) Bioenergy production and forest landscape change in the southeastern U.S. GCB Bioenergy, Accepted.

Coulston JW, Moisen GC, Wilson BT, Fenoy MV, Cohen WB, Brewer CK (2012) Modeling percent tree canopy cover: a pilot study. Photogrammetric Engineering & Remote Sensing, 78, 715–727.

Dale VH, Kline KL, Wiens J, Fargione J (2010) Biofuels: implications for land use and biodiversity. Biofuels and Sustainability Reports: 2014, 1–13. Available at: http://www.esa.org/esa/wp-content/uploads/2014/11/ESA-Biofuels-Report-1.pdf (accessed 7 July 2016).

Dietz J, Jones MB, Stout JC (2010) The impact of biomass crop cultivation on temperate biodiversity. Global Change Biology Bioenergy, 2, 268–309.

Dhondt AA, Wrege PH, Cerritelli J, Sydtenstricker KV (2007) Avian species richness and reproduction in short rotation coppice habitats in central and western New York. Bird Study, 54, 12–22.

Dornburg V, van Vuuren D, van de Ven G et al. (2010) Bioenergy revisited, key factors in global potentials of bioenergy. Energy and Environmental Science, 3, 258–267.

Eckerle KP, Thompson CF (2001) Yellow-breasted chat (Uraeginthus aura). In: The Birds of North America Online (ed. Poole A), Cornell Lab of Ornithology, Ithaca. Available at: http://bna.birds.cornell.edu/bna/species/575 (accessed 11 December 2015).

Eggers J, Troltzsch K, Falacci A et al. (2009) Is biofuel policy harming biodiversity in Europe? Global Change Biology Bioenergy, 1, 18–34.

Energy Independence and Security Act of 2007 (2007) 42 U.S.C. § 17001.

Energy Policy Act of 2005 (2005) 42 U.S.C. § 15801.

ESRI (2012) ArcGIS Desktop. Release 10.1. Environmental Systems Research Institute, Redlands, CA.

Fargione J (2010) Is bioenergy for the birds? An evaluation of alternative future bioenergy landscapes. PNAS, 107, 18745–18746.

Firbank LG (2008) Assessing the ecological impacts of bioenergy projects. Bioenergy Research, 1, 12–19.

Fletcher RJ, Robertson BA, Evans J, Doran PJ, Alalavapatia JRR, Schemske DW (2010) Biodiversity conservation in the era of biofuels: risks and opportunities. Frontiers in Ecology and the Environment, 9, 161–168.

Forsk Consulting LLC (2014) Wood bioenergy US database 2013. Available by subscription.

Fritts SR, Moorman CE, Hazel DW, Jackson BD (2014) Biomass harvesting guidelines affect downed woody debris retention. Biomass and Bioenergy, 70, 382–391.

Fritts SR, Moorman CE, Grodusky SM, Hazel DW, Homyack JA, Farrell CB (2015) Shrew response to variable woody debris retention: Implications for sustainable forest bioenergy. Forest Ecology and Management, 336, 35–43.

Fritts SR, Moorman CE, Grodusky SM, Hazel DW, Homyack JA, Farrell CB, Castleberry SB (2016) Do biomass harvesting guidelines influence herpetofauna following harvests of logging residues for renewable energy? Ecological Applications, 26, 926–949.

Galik CS, Abt RC (2015) Sustainability guidelines and forest market response: an assessment of European Union pellet demand in the southeastern United States. Global Change Biology Bioenergy, 5, 658–669.

Gercky KJ, McKerrow A (2013) Land cover—National inventory of vegetation and land use: U.S. Geological Survey Fact Sheet 2013-3085, 1 p. Available at: http://pubs.usgs.gov/fs/2013/3085/ (accessed 11 December 2015).

Groom MJ, Gray EM, Townsend PA (2008) Biofuels and biodiversity: principles for creating better policies for biofuel production. Conservation Biology, 22, 602–609.

Hammerson G, Jensen J (2004) Lithobates capito. The IUCN Red List of Threatened Species. Version 2015.2. Available at: www.iucnredlist.org (accessed 20 August 2015).

Hartman JC, Nippert JB, Orozco RA, Springer CJ (2011) Potential ecological impacts of switchgrass (Panicum virgatum L) biofuel cultivation in the Central Great Plains, USA. Biomass and Bioenergy, 35, 3415–3421.

Immerzeel D, Verweij P, Hilst F, Faaij AP (2014) Biodiversity impacts of bioenergy crop production: a state-of-the-art review. Global Change Biology Bioenergy, 6, 183–209.

IUCN (2015) The IUCN Red List of Threatened Species. Version 2015.3. Available at: www.iucnredlist.org (accessed 4 November 2015).

Jensen JB, Richter SC (2005) Rana capito, Gopher Frogs. In: Amphibian Declines: The Conservation Status of United States Species (ed. Lannoo MM), pp. 536–538. University of California Press, Berkeley, CA, USA.

Martinuzzi S, Pidgeon AM, Radeloff VC, et al. (2015) Impacts of future land-use changes on wildlife habitat: insights from Southeastern U.S. Ecological Applications, 25, 160–171.

McDonald RJ, Fargione J, Kiesecker J, Miller WM, Powell J (2009) Energy sprawl or energy efficiency: climate policy impacts on natural habitat for the United States of America. PLoS ONE, 4, e6002.

McKerrow AJ, Williams SG, Collazo JA (2006) The North Carolina Gap Analysis Project: Final Report. North Carolina Cooperative Fish and Wildlife Research Unit. Submitted to: The National Gap Analysis Program. U.S. Geological Survey, Biological Resources Division. Available at: http://www.basc.ncsu.edu/nsgap/NCFinal%20Report.pdf (accessed 11 December 2015).

Meehan TD, Hurlbert AH, Gratton C (2010) Bird communities in future bioenergy landscapes of the Upper Midwest. Proceedings of the National Academy of Sciences of the United States of America, 107, 18533–18538.

Mitchell R, Engstrom T, Sharitz R (2009) Old forests and endangered woodpeckers: old-growth in the southern coastal plain. Natural Areas Journal, 29, 301–310.

North Carolina Wildlife Resources Commission (2015) North Carolina Wildlife Action Plan. 1328 p. Available at: http://www.ncwildlife.org/Portals/0/Conserving/
Supporting Information

Additional Supporting Information may be found online in the supporting information tab for this article:

Table S1. Descriptions of the planted pine vegetation classes (‘planted’ or ‘public’) used in our vegetation and species-habitat models.

Table S2. Species-land cover class associations used to model available habitat for 16 wildlife species in NC, USA.

Table S3. A general representation of species-land cover class associations.

Table S4. The conservation listing status of focal species.

Table S5. Ancillary parameters used in species models.

Table S6. The predicted area (thousand ha) of suitable land cover within NC, USA for species from 2010 until 2050 under Costanza et al.’s (2016) bioenergy production scenarios.

Table S7. Definitions of species codes used in Tables S2, S3, and S5.

Figure S1. The predicted distribution of Cerulean Warbler habitat within NC, USA in 2010.

Figure S2. The predicted distribution of Eastern Tiger Salamander habitat within NC, USA in 2010.

Figure S3. The predicted distribution of Gopher Frog habitat within NC, USA in 2010.

Figure S4. The predicted distribution of Mole Salamander habitat within NC, USA in 2010.

Figure S5. The predicted distribution of Oak Toad habitat within NC, USA in 2010.

Figure S6. The predicted distribution of Brown-headed Nuthatch habitat within NC, USA in 2010.

Figure S7. The predicted distribution of Field Sparrow habitat within NC, USA in 2010.

Figure S8. The predicted distribution of Grasshopper Sparrow habitat within NC, USA in 2010.

Figure S9. The predicted distribution of Hairy Woodpecker habitat within NC, USA in 2010.

Figure S10. The predicted distribution of Kentucky Warbler habitat within NC, USA in 2010.

Figure S11. The predicted distribution of Loggerhead Shrike habitat within NC, USA in 2010.

Figure S12. The predicted distribution of Prothonotary Warbler habitat within NC, USA in 2010.

Figure S13. The predicted distribution of Red-headed Woodpecker habitat within NC, USA in 2010.

Figure S14. The predicted distribution of Swainson’s Warbler habitat within NC, USA in 2010.

Figure S15. The predicted distribution of Wood Thrush habitat within NC, USA in 2010.

Figure S16. The predicted distribution of Yellow-breasted Chat habitat within NC, USA in 2010.

Figure S17. Projected trends in habitat availability in NC, USA from 2010 to 2050 under a baseline scenario and five bioenergy scenarios.

Appendix S1. Sources used during the literature review of species’ habitat associations.

Appendix S2. Outline of the geoprocessing steps of habitat map production.