

# 1 Mapping relative forest maturity and stand development for conservation in 2 the conterminous USA

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14

## 15 Abstract

### 16 Context

17 Information on the maturity of forests is important for conservation planning. However, available  
18 information for the USA is inadequate to support national conservation assessment and planning.

### 19 Objectives

20 The main objective was to spatially model at a high resolution the relative level of maturity and stand  
21 development for forests across conterminous USA. A secondary objective was to explore which younger  
22 forests could be attributed to the impacts of severe natural disturbances.

### 23 Methods

24 We modelled the relative level of maturity for forests at a 30 m pixel resolution using spatial data for  
25 forest cover, height and biomass, stratified by forest types and ecoregions. National plot data were used to  
26 validate modelled results. The impact on Young forest from severe wildfire, insects and disease, and  
27 tornados was examined for the years 2000-2019.

### 28 Results

29 Of a total forest area of 248.9M ha, Young forest covered 52.9 M ha (22%); Intermediate 100.3 M ha  
30 (42%); and Mature 86.0 M ha (36%). Results suggest that the modelled data are tracking observed forest  
31 structure and stand development. 1.4 M ha (2.67%) of modelled Young forest was impacted by severe  
32 natural disturbances, with 51.5 (97.33%) M ha of Young forest unimpacted. The distribution of the  
33 disturbance factors varied geographically. The unimpacted Young plus Mature forest are where primary  
34 forests are most likely found.

### 35 Conclusions

36 The forest maturity data can assist forest decision makers in meeting environmental commitments  
37 regarding mitigating forest sector emissions, biodiversity conservation and water quality, including  
38 through prioritizing land for meeting protected area and ecosystem restoration targets.

39 **Key words:** forest maturity, conservation planning, primary forest, old growth, conterminous USA  
40

## 41 **Introduction**

42 Information on the maturity of forests, including forest age, tree species, and stand structure, is critically  
43 important for conservation assessments and planning (Angelstam et al. 2018; Curtis and Gough 2018).  
44 More mature forests, including old growth, are important for biodiversity providing special habitat  
45 conditions and resources (Freund et al. 2015; Frey et al. 2016). Older forests have higher levels of  
46 biodiversity compared to logged forests in every region of the USA (Strittholt et al. 2006; Ducey et al.  
47 2013). In addition, older forests at the stand level generally store more carbon (Moomaw et al. 2019) and  
48 are important for maintaining hydrological processes at the watershed scale (Perry and Jones 2017;  
49 Pypker et al. 2006; Crampe et al. 2021).

50 However, available information for the USA is inadequate to support national conservation assessment  
51 and planning as the relevant data sets are either at too coarse a spatial resolution or do not reflect recent  
52 forest condition (Pan et al. 2011). Conservation requires information about the condition of different  
53 forest ecosystems including distinguishing between those that are primary forests - largely undisturbed by  
54 modern land uses and dominated by ecological processes including successional stages arising from  
55 natural disturbance processes (Kormos C.F. et al. 2017) - and secondary and regrowth forest which are  
56 the result of human management including logging. We address this information gap by drawing upon  
57 available spatial data to provide a wall-to-wall high resolution (30m) spatial dataset of relative forest  
58 maturity and stand development for the conterminous USA.

59 Forest at the older end of the stand development continuum have become increasingly rare due to land use  
60 impacts, mainly logging for commodity production that has shifted forest stand development to perpetual  
61 early re-growth phases (20-60 years depending on growth rates) (Foley et al. 2009). Severe wildfires also  
62 impact forest age and stand development (Pan et al. 2011), with consequences to structure and  
63 composition varying with burn severity (Reilly et al. 2018; Reilly et al. 2020). In the western USA, an  
64 increase in extreme fire weather conditions and area burned at high severity has now been attributed to in  
65 part to human influenced climate change (Abatzoglou and Williams 2016; Parks and Abatzoglou 2020).

66 Stand development is also impacted by insect and disease infestations (Fei et al. 2019) and tornados.  
67 Forest age and level of stand development are measured through tree ring analysis (e.g., core drill samples  
68 from living trees) or models based on measurements of forest structure – canopy height and density.  
69 Other forest characteristics indicative of stand age include vertical vegetation layering and coarse woody  
70 debris. Differences in the longevity, life history traits and niche requirements of tree species means that in  
71 many ecosystem types, the taxonomic composition of the dominant canopy species can reflect seral stages  
72 progressing from pioneer to late successional (Huston and Smith 1987). Gap-phase dynamics are  
73 diagnostic of older forests that result when some of the dominant trees die and provide opportunities for  
74 release of understory vegetation that fill those gaps overtime (for examples in the Pacific Northwest, see  
75 (Spies 2004); in eastern forests see (Davis et al. 1996). The lack of a large-scale disturbance over decades  
76 allows forest succession to proceed through mature stages. Furthermore, environmental factors that  
77 regulate plant growth, ecosystem processes rates and site productivity – thermal, moisture, radiation and  
78 nutrient regimes – also result in variation within the ecosystem type of forest structure in terms of  
79 canopy, height, canopy density, and above ground woody biomass, further confounding the use of  
80 structure to infer age (Shao et al. 2018).

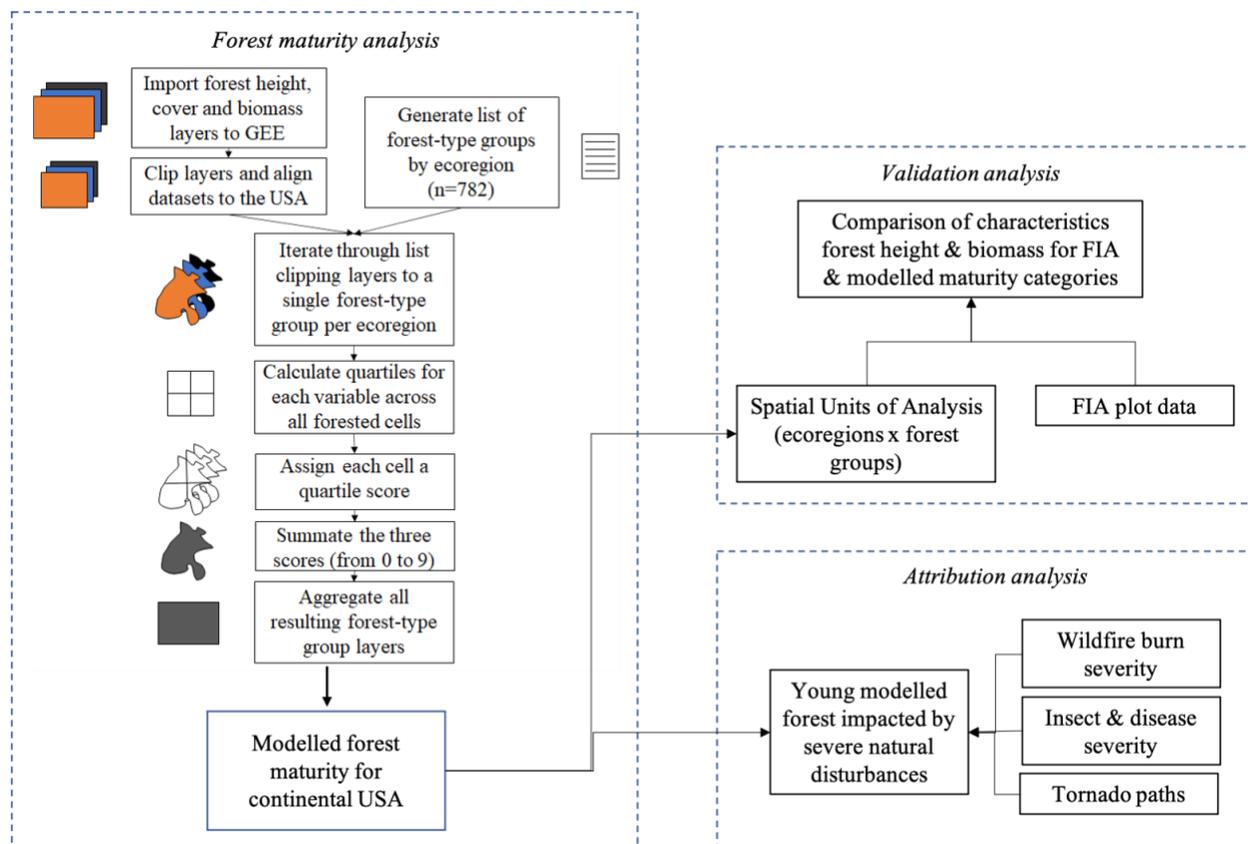
81 Our main objective was to map the relative level of maturity and stand development for forests across  
82 conterminous USA, including both naturally regenerating forests and plantations. A secondary objective  
83 was to explore which younger forests could be attributed to the impacts of severe natural disturbances.

## 84 **Methods**

### 85 **Overview**

86 Our approach to mapping stand development was limited by the need to draw upon publicly available  
87 spatial data sets with wall-to-wall coverage for conterminous USA. This meant that suitable spatial data  
88 were available for only a subset of stand development characteristics. We mapped the relative level of

89 forest maturity and stand development using three published spatial data sets for forest cover, height and  
 90 biomass derived from modelled satellite data (Table 1). These data were stratified by USA Ecoregions  
 91 Level III (n=28) (Omernik and Griffith 2014) and Forest Types Groups (n=85) (Ruefenacht et al. 2008) to  
 92 account for the influences of variation in life history traits governing tree longevity and local  
 93 environmental conditions on plant growth and ecosystem processes, as well as differing human and  
 94 natural disturbance regimes. We used field measurements from the Forest Inventory and Analysis plot  
 95 database (FIA 2022) to compare with our modelled forest maturity map as a form of validation. We used  
 96 a time series of available spatial data to examine how modeled Young forests has been impacted by  
 97 severe natural disturbances. An overview of the analytical workflow is given in Fig. 1. Further details on  
 98 the methodology are provided below and in Supplementary Information - Methods.



99

100 **Fig. 1** Workflow showing main steps in the calculation of the forest maturity model for conterminous USA, the  
 101 validation analysis and the wildfire attribution analysis. The three 30m resolution spatial data sets for forest cover,

102 height and biomass were analyzed within 872 spatial units of analysis (SUA) defined by the intersection of  
 103 ecoregions and major forest types. FIA plot data were used for a validation analysis. A time series data set was  
 104 assembled to identify which areas of modelled Young forest had been severely impacted by wildfire, insects and  
 105 disease, and tornado paths. Further details in Supplementary Information – Methods

106 **Expert workshops**

107 A series of regional workshops were held via zoom to consult with ecological and conservation forest  
 108 experts. The workshops were conducted between September and November 2021. In total over 40  
 109 experts attended the workshops with each focused on a major forested regions within the conterminous  
 110 USA. Key objectives were to obtain expert advice on: (a) the most appropriate level of ecoregion and  
 111 forest ecosystem classification to use for a continental analysis; (b) potential sources of error and  
 112 limitation in these data; (c) the suitability of available forest data sets for modelling and analysis; (c)and  
 113 (d) to provide feedback on preliminary modelling results. Further details on the workshop methods and  
 114 outcomes are provided in Supplementary Information – Methods.

115 **Forest maturity model**

116 The three spatial data layers of forest cover, height and aboveground living biomass were available for the  
 117 conterminous USA (Table 1). Spatial analyses were undertaken using Google Earth Engine (Gorelick et  
 118 al. 2017). As the three data layers were generated using the Global Land Analysis and Discovery’s  
 119 (GLAD) Landsat Analysis Ready Data (ARD), they shared the same 30m pixel resolution.

120 Table 1. Details for the spatial data layers used in the forest maturity modelling and the attribution and  
 121 validation analyses.

Layer	Description	Data type and scale / resolution	Calibration data / validation approach	Source
Forest cover	Percent tree cover stratum (e.g. >50% crown cover to ~0% crown cover)	Raster (30m)	High resolution imagery	(Hansen et al., 2013) (Hansen et al. 2013) updated to 2010 (GLAD)

Forest height	Forest canopy height	Raster (30m)	Vegetation structure data collected using airborne lidar instruments (ALS) and GEDI field plots	(Potapov et al., 2021)
Forest biomass	Modelled estimates of aboveground living biomass	Raster (30m)	Based on machine learning of satellite band ratios, plot measurements of biomass, environmental variables	(Harris et al., 2021)
Ecoregions (Levels III)	Areas of similar ecosystems	vector data layer (at or above 1:24,000 scale)	Field verification trips across 30 US states.	(Omernik and Griffith, 2014)
Forest Type Groups	Aggregation of forest types into 28 categories	Raster (250m)	FIA inventory plot data; spatial environmental data layers	(Ruefenacht et al., 2008)
Burn severity	Fire severity using variants of the normalized burn ratio (NBR, dNBR and RdNBR)	Raster (30m)	Calibrated per fire using high resolution imagery for both pre and post fire	(Eidenshink et al., 2007)
Insect and disease	Areas of forest with insect and disease outbreaks	Polygonised from 240 m resolution raster	Aerial and ground surveys	(USFS, 2022)(Frank J.K. Jr. et al. 2014)
Tornados	Tornado paths	Spatial vector data layer (scale not reported)	Tornado reports later field validated by local National Weather Service forecast offices	(NOAA 2020)

122

123 Fig. 1 gives an overview of the workflow to create a seamless conterminous-USA wide forest maturity  
124 spatial data layer. We created a spatial vector file of each major forest group for each Level III  
125 Ecoregions (Omernik and Griffith 2014). Spatial data layers were generated based on spatial coverage for  
126 the Major Forest Groups found in each ecoregion, resulting in a total of 782 unique forest group-  
127 ecoregion combinations. Quartile values for the three forest variables (canopy cover, height and biomass)  
128 were calculated for each pixel within each of the 782 combinations. A score was then calculated for each  
129 pixel as follows: (a) the lowest quartile value for each metric was given a score of 0 and the highest a  
130 score of 4; then (b) the four metric scores were summed giving a range in possible values from 0 (lowest  
131 quartile for the three variables) to 9 (highest quartile for the three variables), representing 10 ordinal  
132 forest maturity classes. Based on expert feedback from the workshops, we then produced a simplified  
133 maturity map by classifying pixels with a score of 0 as “indeterminant, those with scores of 1-3 as

134 “Young”, scores 4-6 “Intermediate” and scores of 7-9 as “Mature”. Using a global spatial data set (WRI  
135 2016), we analyzed the modelled forest maturity map to identify how much of each maturity class was  
136 plantation rather than naturally regenerating forest. For further see Supplementary Information –  
137 Methods.

### 138 **Natural disturbance attribution analysis**

139 We assembled a 20-year time series for the period 2000-2019 from available spatial data sets for the  
140 conterminous USA to examine the extent to which modelled young forest have been subject to severe  
141 impacts from natural disturbances: high fire severity (Finco M. et al. 2012); very severe insect and disease  
142 outbreaks (USFS 2022) ; and high winds within tornado paths (Harris et al. 2016; NOAA 2022). The total  
143 area of modelled Young forest impacted by each severe disturbance factor was calculated. Further details  
144 are provided in Supplementary Information - Methods

### 145 **Validation analysis**

146 We used FIA plot data as an independent data source for validating the modelled forest maturity map. For  
147 this validation analysis, the spatial units of analysis (SUA) for comparison with the FIA plot data were  
148 generated from the intersection of the map of 85 USA Ecoregion Level III with the maps of the 28 Forest  
149 Type Groups. Those SUAs were analyzed for which there were at least 10 FIA plots for each of the three  
150 FIA Structural Stage Classification levels (Pole, Mature, Late) (n=41). For each of these 41 SUAs, we  
151 calculated aggregate statistics from the quartiles and median values for canopy height and biomass from a  
152 random sample of pixels within each of the three modelled maturity levels (Young, Intermediate, Mature)  
153 with 1.5-5% of pixels sampled (Table 2). We assumed that the FIA maturity levels (Pole, Mature, Late)  
154 were sufficiently equivalent with the modelled Young, Intermediate and Mature categories, respectively,  
155 to support comparisons among them. Further details are provided in Supplementary Information -  
156 Methods.

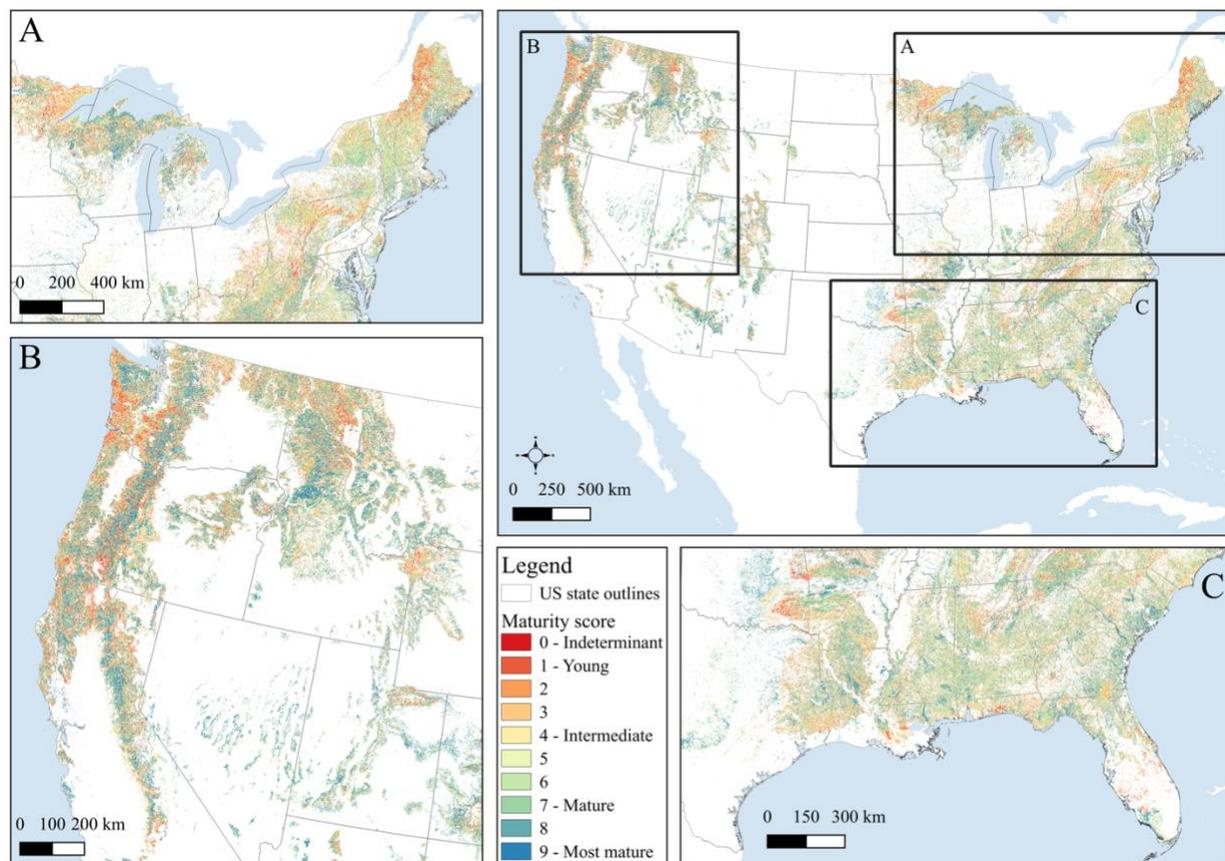
157 Table 2. Characteristics of the sampled FIA plot and modelled forest maturity pixel data used in the  
 158 validation analysis. Details are provided here for the six SUA with largest forest cover.

			Modelled forest data	FIA plot data	
<i>Spatial Unit of Analysis (Major Forest Group x Ecoregion Level III)</i>	<i>total forest area</i>	<i>plantations (% of total forest area)</i>	<i>total area of forest sampled (ha); forest area sampled (%)</i>	<i>no. biomass plots</i>	<i>total area sampled by biomass plots (ha); forest area sampled (%)</i>
Northeastern Highlands-Maple/Beech/Birch Group	10,262,450	0.01	513,122 (5)	1,480	100 (0.0010)
Southeastern Plains-Loblolly/Shortleaf Pine Group	8,704,181	49.1	435,209 (1.5)	146	10 (0.0001)
Ridge and Valley-Oak/Hickory Group	5,624,025	2.16	281,201 (5)	668	45 (0.0008)
Northern Rockies-Douglas-fir Group	4,717,516	8.64	235,876 (5)	425	29 (0.0006)
Southern Coastal Plain-Longleaf/Slash Pine Group	3,813,097	42.04	190,655 (5)	440	30 (0.0008)
Coast Range-Douglas-fir Group	3,284,700	27.8	164,235 (5)	558	38 (0.0011)

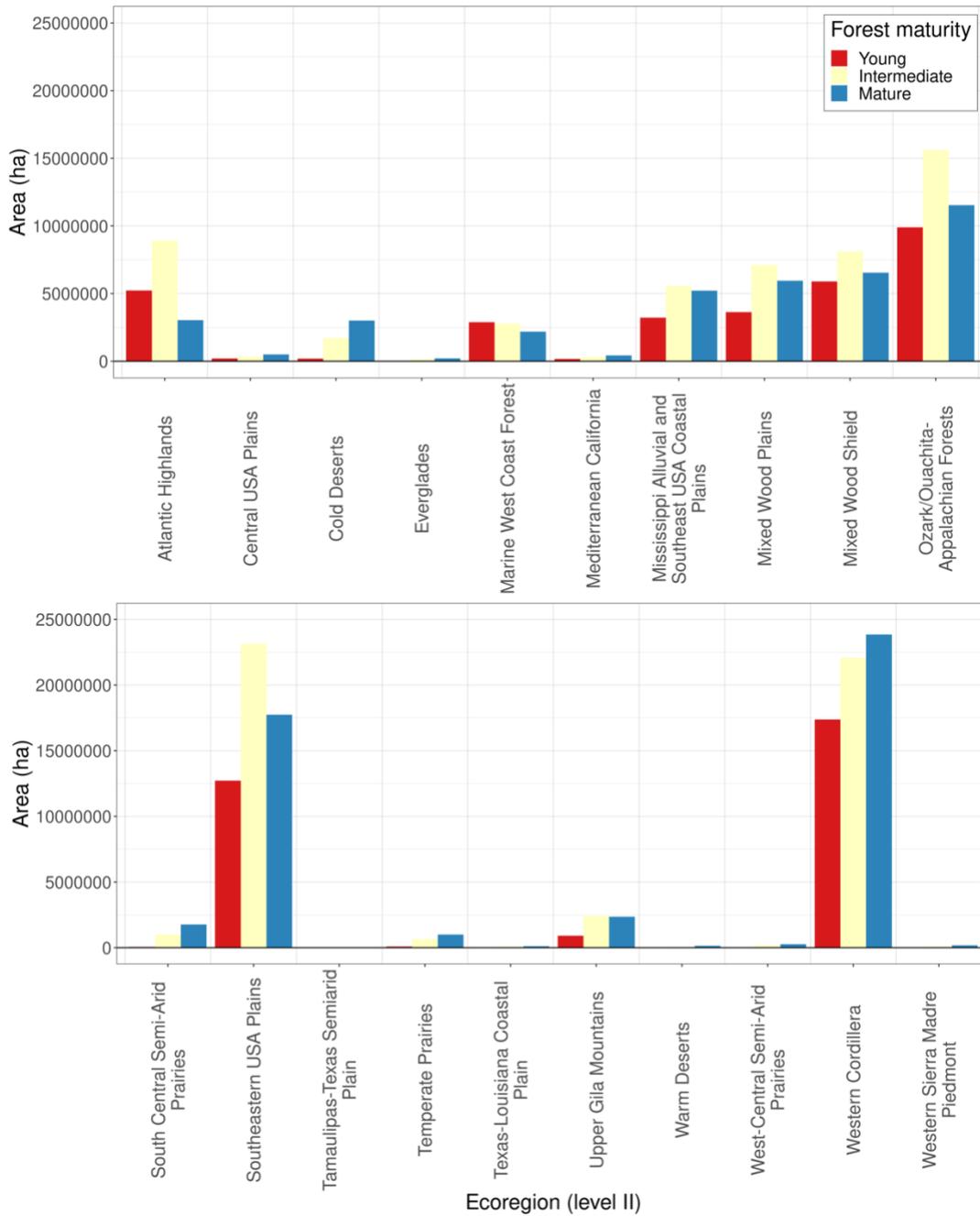
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160 **Results**

161 The forest maturity map comprised 10 ordinal classes (Fig. 2). These were grouped into Indeterminant  
 162 (class 0), Young (classes 1-3), Intermediate (classes 4-6) and Mature (classes 7-9). Total forested area  
 163 was 248.9M ha categorized by Young (52.9M ha; 22%); Intermediate (100.3M ha; 42%); and Mature  
 164 (86.0M ha; 36%). Maturity levels varied among ecoregions as expected with most regions showing  
 165 Mature forests in a matrix of Young and Intermediate. Only two ecoregions had more Mature forest,  
 166 including the most forested ecoregion (Western Cordilla). However, all forested ecoregions still retain  
 167 substantial areas of Mature forest (Fig. 3). The total area of plantation in the conterminous USA is  
 168 24.10M ha of which most was Intermediate (11 M ha, 46%) followed by Mature (7 M ha, 29%), and  
 169 Young (6 M ha, 25%). Note that we used Ecoregions Level II (rather than Level III) in Fig. 3 simply to  
 170 provide a high-level overview of these results.



172 **Fig. 2** Forest maturity ranking. High maturity ranking score means a pixel has denser forest cover, a taller  
 173 canopy and larger biomass compared to a younger forest. Note that this is a relative index  
 174



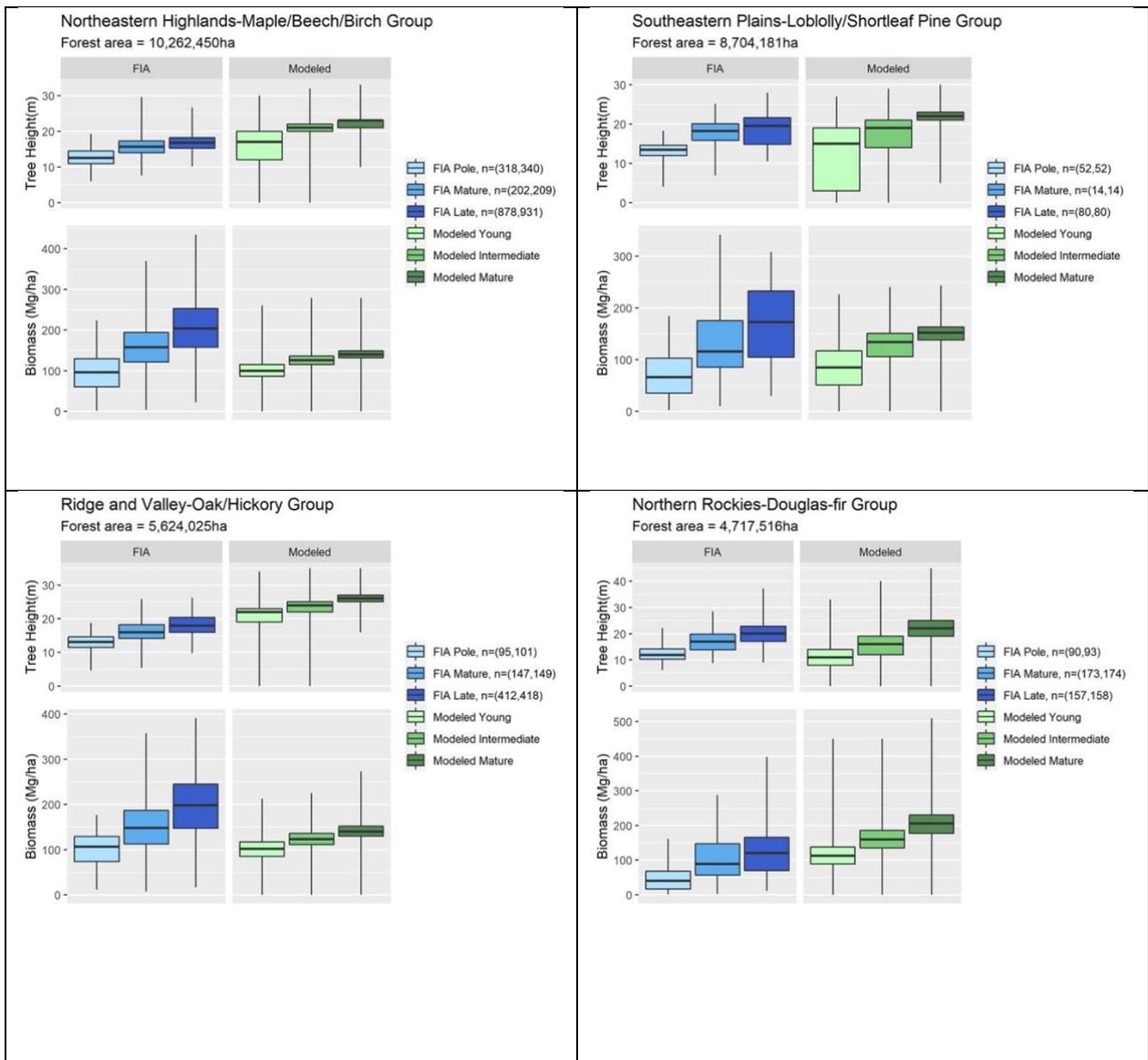
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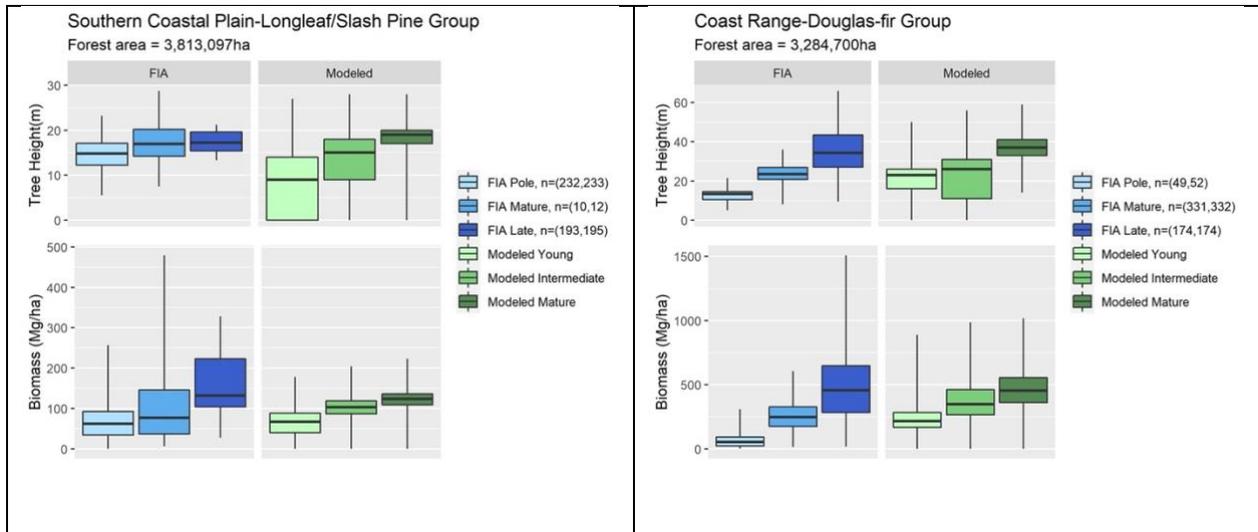
177 **Fig. 3** The area (ha) of young, intermediate and mature forest maturity within USA Ecoregions Level II.  
 178 The three categories of maturity are based on the 10 ranked forest maturity classes (Fig. 2): Young 1-3;  
 179 Intermediate 4-6; Mature 7-9. A percent area version of the figure can be found in the Supplementary  
 180 Information Fig. S11)

181

182 Visual inspection of the box plots for the six SUA with the largest forest area show that the canopy height  
 183 and biomass increase with forest maturity level in both the plot and modelled data. It is also evident that  
 184 there is considerable agreement between the FIA plot data and modeled forest maturity as the values for  
 185 the modelled data largely fall within the range of the plot data, and the median values are reasonably  
 186 aligned (Fig. 4). However, in some cases the ranges differ considerably between them (plots for all 41  
 187 SUA are provided in Supplementary Information – Figures, Fig. SI3).

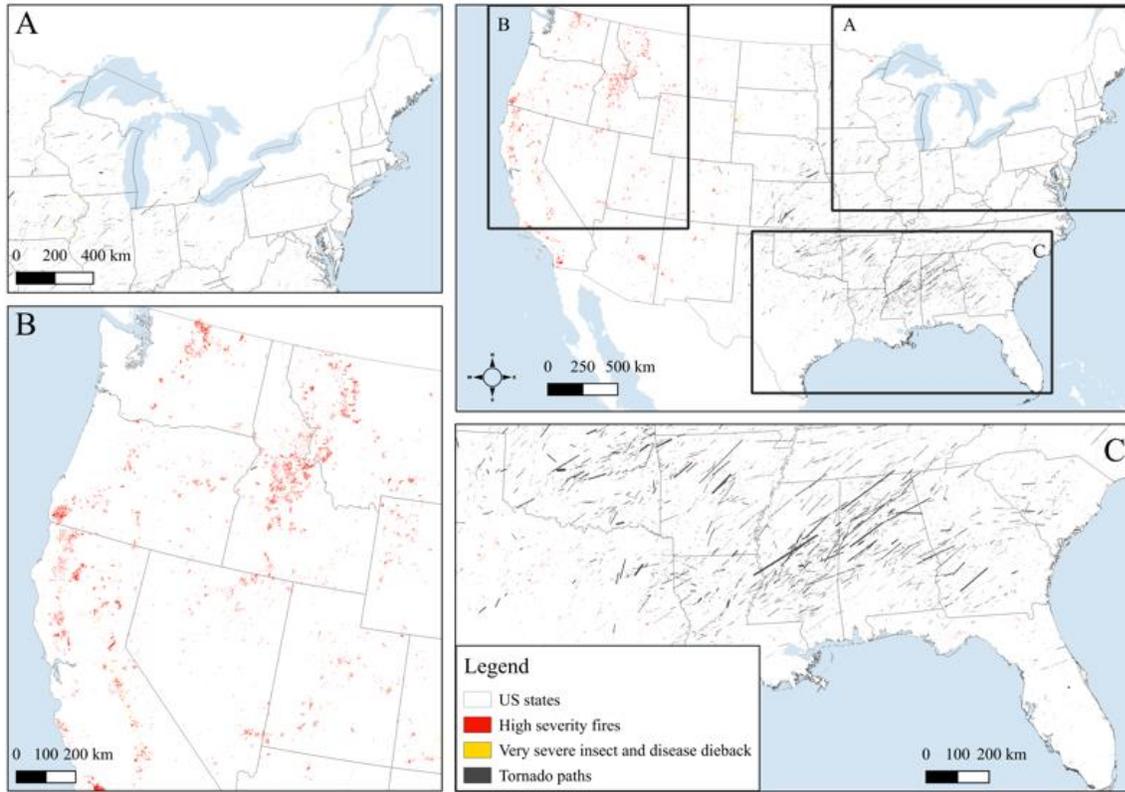
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189 **Fig. 4** Comparison of characteristic canopy height and biomass from modelled forest maturity with  
 190 estimates from FIA plot data. The spatial units of analysis (SUA) were the intersection of USA  
 191 Ecoregions Level III with modelled geographic distributions of major forest groups. SUA were analyzed  
 192 that contained at least 10 FIA plots for each of the four FIA maturity levels. The SUA are labelled  
 193 according to their major forest group name. Box plots defined by interquartile range (first quartile,  
 194 median, third quartile) with whiskers being minimum and maximum based on the most recent values  
 195 from all available FIA plots within each of the three stand structure categories (Pole, Mature, Late), for  
 196 each SUA

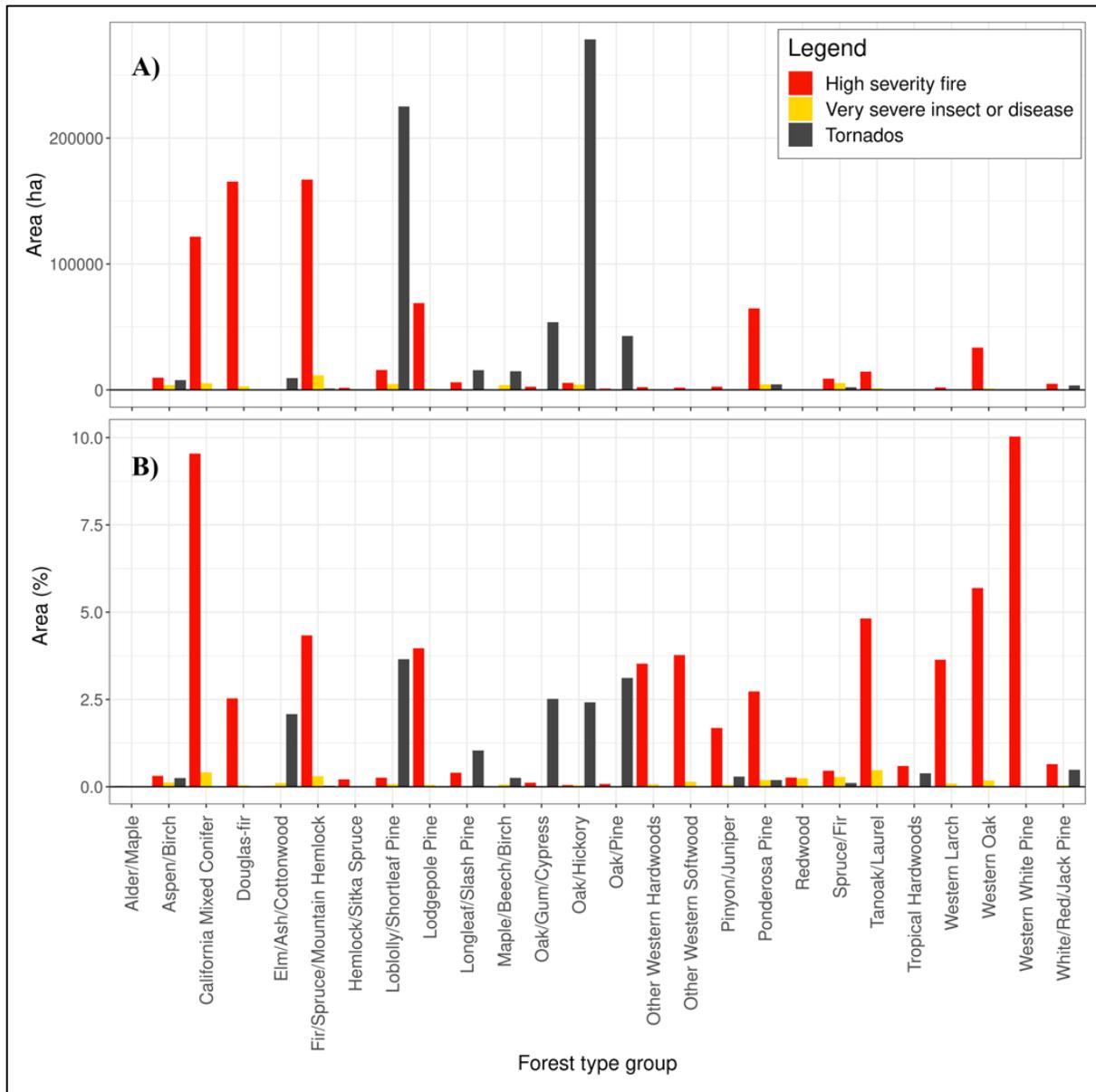
197 The spatial distribution of the three natural disturbance factors (wildfire, insects and disease, tornado  
 198 paths) varied geographically with wildfires most prevalent in the Northwest and tornados in the Southeast  
 199 (Fig. 5). It follows that the correlation between these factors and Young forest varied among Forest Type  
 200 Groups (Fig. 6): California Mixed Conifer and Western White Pine Groups have been most impacted by  
 201 wildfire; Loblolly/Shortleaf Pine and Oak Pine Groups most by tornados; and Fir/Spruce/Mountain  
 202 Hemlock and California Mixed Conifer Groups by insects and disease.



203

204 **Fig. 5** Spatial distribution of the three natural disturbance factors (wildfire, insect and disease, tornado  
 205 paths)

206



207

208 **Fig. 6** The area (a) and percentage areas of modelled young forest in Forest Type Group that have  
 209 experienced severe impacts from wildfire, insect and disease or tornadoes during the period 2000-2019

210 A total of ~1.4 M ha of modelled Young forest was impacted by high burn severity wildfires, severe  
 211 insects and disease, or tornado paths, though this is only 2.67% of the total area (Table 3). This leaves  
 212 ~51.5 M ha of Young forest which has not been impacted by these natural disturbance during the 20-year  
 213 time period examined here.

214 Table 3. The area and percentage of modelled Young forest impacted by high burn severity wildfires,  
 215 severe insects and disease, and tornado paths.

High severity fire		Very severe insect and disease		Tornado paths		Total impacted by natural disturbances		Unimpacted Young forest	
area (ha)	%	area (ha)	%	area (ha)	%	area (ha)	%	area (ha)	%
701,824	1.33	51,931	0.01	660,094	1.25	1,413,849	2.67	51,494,949	97.33

216

217

218 **Discussion**

219 To our knowledge, this is the first map of forest maturity classes at 30m resolution for the conterminous  
 220 USA. Our results complement earlier studies that mapped forest aged distributions at a 1-10 km  
 221 resolution (Besnard et al. 2021; Pan et al. 2011), recent studies examining the impact of natural  
 222 disturbances on the net carbon exchange of forested lands (Harris et al. 2021), and the contribution of  
 223 forested lands to meeting conservation targets (Law et al. 2021). With growing interest in the ecological  
 224 and conservation values of older forests in natural climate solutions (DellaSala et al. 2020; Griscom et al.  
 225 2017; Law et al. 2021; Mackey et al. 2020; Moomaw et al. 2019). information on mature forests is  
 226 relevant to conservation policy makers. However, it is important that the assumptions and limitations of  
 227 the data and approach are understood for the forest maturity map to be applied appropriately.

228 We assumed that for a given Forest Type Group in a given ecoregion, the level of maturity would be  
 229 monotonically related to increasing cover, height and biomass. An initial visual inspection of the  
 230 modelled forest maturity map identified two landscape settings where the forest was likely erroneously  
 231 assigned a younger forest maturity class. One was forests bordering the alpine zone that naturally have a  
 232 sparser and shorter canopy and support smaller biomass stocks forests compared to a similar type at a  
 233 lower elevation. Less obviously, are forests in climatically drier ecoregions on exposed topographic

234 positions that naturally would be sparser, shorter and have less biomass than similar forest types nearby  
235 with higher site productivity (McKenney and Pedlar 2003). The Oak/Hickory Forest Type Group also had  
236 some anomalous results with lower-than-expected areas of Young forest. This is likely the result of  
237 substantial wildfire suppression in these fragmented forests across their range (Nowacki and Abrams  
238 2008). In addition, closer inspection of the input data revealed differences in how non-forest land was  
239 mapped. The Landfire tree cover layer used to delineate Alpine vegetation has a stricter classification of  
240 forest than the other inputs. We also tacitly assumed that the quantile thresholds correspond to age  
241 classes. However, it could be that an ecoregion with a majority of mature forests could have some mature  
242 mapped as intermediate/young, while the opposite could occur in an ecoregion with a majority fraction of  
243 young forests. This could also explain why substantial areas of plantations were mapped as having mature  
244 forest even though they are managed to be harvested at a young age.

245 We compared the modelled forest maturity with FIA plot data. The validation results suggest that the  
246 modelled forest maturity map is realistically tracking forest structure and stand development. However,  
247 the categories of stand development (FIA – Pole, Mature, Late; Modelled – Young, Intermediate, Mature)  
248 were derived using different methods and therefore can only be considered indicatively equivalent. While  
249 overall the box plots frequently share overlapping inter-quartiles and medians, there are interesting  
250 difference, for instance in cases where the plot data samples a wider range of biomass than the modelled  
251 data. These differences warrant further investigation.

252 From a conservation perspective, primary forests in the temperate biome are an increasingly rare asset  
253 globally (FAO 2020; Mackey et al. 2015). Our attribution analysis revealed the areas of modelled Young  
254 forest that have been subjected to at least one of the severe natural disturbances examined (wildfire,  
255 insects and disease, tornado paths) between the years 2000-219. Regardless of their level of stand  
256 development, natural forests can be defined as “primary forest”, so long as the impacts that have resulted  
257 in younger rather than older maturity levels are the result of natural disturbances rather than human land  
258 use (Kormos C.F. et al. 2017). It is reasonable to assume therefore that the (non-plantation) forest

259 impacted by severe natural disturbances could be primary forest, so long as they have not also been  
260 impacted by logging prior to the year 2000. Similarly, primary forest is also likely to be found in the  
261 Mature forest category where it would be typically recognized as old-growth forest. However, confirming  
262 more accurately the location of the remaining primary forest in conterminous USA requires further  
263 attribution studies using logging and land use history data. Unfortunately, a consistent spatial logging data  
264 set is not currently available for the conterminous USA.

265 A limitation of our modelled forest maturity is that it does not directly provide a measure of forest stand  
266 age. However, as noted, our maturity levels (Young, Intermediate, Mature) are analogous with the FIA  
267 Structural Stage Classification levels (Pole, Mature, Late), and the validation analysis comparing the FIA  
268 plot data with our modelled data suggest they are indicative of age classes typically used in forest  
269 classifications. In any case, assigning an age to forest development is not straightforward except where  
270 field-based measurements are available or in the case of an even-aged forest area following a complete  
271 stand-altering natural disturbance or land use. Forest age and the level of stand development reflects the  
272 influence of multiple interacting factors, including the longevity of dominant canopy tree species, the  
273 physical environmental factors that regulate plant growth, land use including logging, and the impacts of  
274 natural disturbance regimes. As tree age reflects genetically determined life history traits, and phenotypic  
275 responses to site conditions, the characteristic range varies among and between forest ecosystem types, as  
276 well as physical environmental conditions. Some of the Major Forest Groups in the East and Southeast  
277 have forest cover that is either recovering from relatively recent clearing and logging or are plantations  
278 with a timber harvesting regime that maintains a young age structure. It follows that in these forests, the  
279 Mature class may indeed be the oldest forest but are still young relative to the age these trees would reach  
280 under natural conditions. This is probably also the case with the plantation results as they are managed to  
281 be harvested at a relatively young age and therefore the most mature stands will be young compared to a  
282 mature natural forest.

283 The Forest Type Groups, stratified by USA Ecoregions Level III, were used to represent the major  
284 differences in forest ecosystems. However, as these Groups are only intended to indicate broad  
285 distribution patterns of forest cover in the USA, they represent a highly generalized level of ecological  
286 organization within which resides a rich forest biodiversity. Furthermore, the spatial models (i.e., maps)  
287 of Forest Types Groups used here are based on statistical models correlating FIA plot data with  
288 environmental and remotely sensed variables, with an overall accuracy of 65% (Ruefenacht et al. 2008).  
289 Also, it should be noted that using these maps for spatial stratification purposes means there was a minor  
290 element of circularity in our validation analysis in that the FIA plot data were input data to the modelled  
291 potential distribution of the Forest Type Groups. These were used to stratify forest cover in calculating  
292 the aggregate statistics shown in the box plots (Fig. 4). Future research could investigate using the general  
293 approach presented here but applied to the next level of Forest Type mapping that provides maps for 141  
294 forest types, or to other data sources with more finely defined forest ecosystem types and mapping for  
295 particular ecoregions.

296 Other than calculating aggregate statistics (see Table 2), we did not distinguish in this analysis between  
297 naturally regenerated forests and plantations. However, to apply the modelled forest maturity mapping for  
298 conservation assessment purposes, this distinction is critical. Plantations are a significant component of  
299 the USA forest cover. Three of the largest forested SUA support significant areas of plantations (Table 2)  
300 including the Southeastern Plains-Loblolly/Shortleaf Pine Group (49%) and the Coast Range-Douglas-fir  
301 Group (28%). Conservation applications of the modelled forest maturity map will also require further  
302 partitioning by land ownership and management categories.

### 303 **Conclusion**

304 Further research is needed to calibrate the relative forest maturity mapped here with actual forest age  
305 classes, undertake analyses at a finer level of forest ecosystem delineation, and to investigate the impact  
306 of logging on, and its interactions with, the natural disturbances impacting on forest stand development

307 and the distribution of modelled Young forest. While recognizing the limitations of the methods and data  
308 used here, the modelled forest maturity map and related analyses can make a useful contribution to forest  
309 planning and management by providing information needed by decision makers to meet national and  
310 international environmental commitments regarding mitigating greenhouse gas emissions from the forest  
311 sector, biodiversity conservation and the provision of clean water. The new high resolution spatial data on  
312 the level of forest maturity can be used to prioritize land for conservation through, for example,  
313 government programs aiming at meeting protected area targets. Information on the likely causes of Young  
314 Forest can assist planning for ecosystem restoration, for example, by identifying areas where stand  
315 development is the result of human land use pressures and a change in forest management is required.  
316 However, these applications will require further research to examine the intersection of forest maturity  
317 with spatial data related to the density of forest carbon stocks, water quality and flow, and the habitat  
318 suitability and presence of threatened and characteristic wildlife species.

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## 419 **Statements & Declarations**

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## 422 **Competing Interests**

423 The authors have no relevant financial or non-financial interests to disclose.

## 424 **Author Contributions**

425 All authors contributed to the study conception and design. Material preparation, data collection and  
426 analysis were performed by Brendan Mackey and Heather Keith. The first draft of the manuscript was  
427 written by Brendan Mackey and all authors commented on previous versions of the manuscript. All  
428 authors read and approved the final manuscript.

429 **Data Availability**

430 The datasets generated during and/or analysed during the current study are available from the  
431 corresponding author on reasonable request.

432 **Supplementary Information Files**

433 *Supplementary Information - Methods.* This file contains additional details on the methods in relation to:  
434 data assembly and visualization; the expert workshops; forest maturity cluster analysis; as well as the  
435 validation and attribution analyses

436 *Supplementary Information – Figures.* This file contains three supplementary figures. Figure SI1 is a  
437 screen shot of one of the interacting mapping apps developed for the expert workshops. Figure SI2 shows  
438 the percentage area of young, intermediate and mature forest within USA Ecoregions Level II. Figure SI3  
439 presents the box plots for all the Spatial Units of Analysis (SUA) that compare the canopy height and  
440 biomass from the modelled forest maturity with estimates from the Forest Inventory and Analysis (FIA)  
441 plot data

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