

1 **Enhancing the provisioning of ecosystem services in South Korea**  
2 **under climate change: The benefits and pitfalls of current forest**  
3 **management strategies**

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22  
23 Length of the manuscript: 7697, Number of Figures: 1, Number of Tables: 0

## Abstract

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We present a study about the integrated impact of adaptive management and climate change using spatially-explicit tools at first time for assessing changes in forest ecosystem services in South Korea. The aims of this study were to: (i) project potential changes among forest ecosystem services under various scenarios, (ii) assess the impact of forest management [Sc1: controlled, Sc2: business as usual, and Sc3: adaptive management] and climate change through comparison among them, and (iii) find insight for strategies to implementing both sustainable society and environment. For this purpose, the integrated tool was applied to analyze the changes in five ecosystem services: forest carbon storage, carbon sequestration, industrial wood production, freshwater supply, and forest recreation. The simulated results show that the ratio of these five ecosystem services of Sc3-to-Sc2 in the 2040s was estimated as 88.4%, 114.7%, 105.4%, 228.2%, and 86.5%, respectively. These results showed clear trade-offs between industrial wood production and freshwater supply on one side and forest recreation and forest carbon storage on the other side. In the case of carbon sequestration, the harvest activity might provide a negative effect in the short term, but it with a longer-term perspective could be positive through reforestation activity. In addition, this study showed that future climate change until 2050 in Korea could have a generally negative influence on forest carbon sequestration but that these negative effects could be partly offset through harvest management. Therefore, the plans of the spatial distribution of management activities for the equilibrium of demand and supply in ecosystem services should be required.

**Keywords:** *industrial wood production, carbon storage, recreation area, fresh water supply, carbon sequestration, forest harvest*

## 46 **1. Introduction**

47 Today, the value and usefulness of forests are increasing (Bonan 2008). Forest  
48 ecosystems are increasingly understood as multifunctional production systems that provide  
49 many products and services, such as timber, various non-wood products, recreational amenities,  
50 habitats, protected water resources, erosion control, and carbon sequestration (Pang et al. 2017).  
51 Forest ecosystem services are significantly affected by management activities such as  
52 afforestation, thinning, and harvest activities (Kim et al. 2019; Bugmann et al. 2017). Because  
53 ecosystem services are typically highly interlinked, optimizing one can affect others (Bennett  
54 et al. 2009); in other words, these interactions can lead to trade-offs (negatives interactions) or  
55 synergies (positives interactions). For example, afforestation reduces soil erosion (Vicente-  
56 Vicente et al. 2019) and increases carbon sequestration (Kim et al. 2016) and soil organic  
57 carbon (Vicente-Vicente et al. 2019); however, increased forestland might lead to reduced  
58 water yield, especially in semi-arid regions (Cao et al. 2010). Relationships among multiple  
59 ecosystem services can be analyzed and understood through trade-offs and synergies, which  
60 have become important topics in the study of ecosystem services at the global and regional  
61 scales (Ray et al. 2015).

62 Quantifying and exploring the spatiotemporal characteristics of ecosystem service  
63 trade-offs and synergies are beneficial to ecosystem management because they can mutate over  
64 time (Tilman 2000); some may appear immediately following ecosystem service changes (Kim  
65 et al. 2016), but some might take much longer to become apparent (Cao et al. 2010). Trade-  
66 offs arise when more of a particular ecosystem service is captured by one stakeholder at the  
67 expense of others, or when the provision of one ecosystem service is reduced as a consequence  
68 of increased use of another ecosystem service (Rodriguez et al. 2006). Such changes could be  
69 the result of explicit choices or arise without premeditation or awareness. Forests can be

70 sensitive to climate change because the long life-spans of trees even though they have adaptive  
71 capacities such as high phenological plasticity to against future environmental changes (Seidl  
72 et al. 2017; Vitasse et al. 2010). Associated with climate change, factors that affect forest  
73 ecosystems can act independently or in combination (Lindner et al. 2014). Therefore, forestry  
74 researchers should know that ecological factors related to climate change as well as to  
75 manmade environmental changes can change the synergies and trade-offs among forest  
76 ecosystem services (Lindner et al. 2014; Rodriguez et al. 2006).

77         At present, one of the major management issues faced by the Korea Forest Service  
78 (KFS) and the country's forest sector is understanding the impacts of climate change and  
79 optimizing the use of forest resources to meet national Sustainable Development Goals (SDGs)  
80 (KFS 2017). Recently, the KFS lowered the legal final cutting age and is planning to increase  
81 the annual harvested area from 15,000 ha in the 2010s to 35,000 ha in the 2030s in order to  
82 increase timber production and improve the imbalance in forest age-class distribution (KFS  
83 2017). Therefore, it is necessary to project corresponding changes in ecosystem services under  
84 alternative forest management scenarios to provide decision-making support regarding  
85 integrated sustainability assessment of policy and planning alternatives.

86         This study aimed to project potential changes among forest ecosystem services under  
87 various scenarios that were related to alternative forest management policies and climate  
88 change. Toward this aim, we simulated six scenarios based on three forest management  
89 strategies and two future climate conditions for the whole forest area in South Korea. In this  
90 study, we present models to assess the project potential changes among forest ecosystem  
91 services. We used five ecosystem services, C storage (CS), C sequestration (Cse), freshwater  
92 supply (WS), industrial wood production (IWP), and forest recreation (RE), to derive key  
93 indicators of trade-offs and synergies in forest ecosystem services. Our study can enhance the

94 comprehension of complex interactions among multiple ecosystem services under different  
95 management and climate change scenarios.

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## 97 **2. Materials and methods**

### 98 **2.1. Study area and base map of Korean forests**

99 South Korea is located at 124° 54' – 131° 06' E and 33° 09' – 38° 45' N (Fig. S1a).  
100 Forest area covers 63.7% (6,369,000 ha) of the total land area, and evergreen, deciduous broad-  
101 leaved, and mixed forests occupied approximately 40.5%, 27.0%, and 29.4% of the total forest  
102 area in 2018, respectively (KFS 2019 – Fig. S1c). We used the following spatial data sets in  
103 this study: Landsat TM and Spot remote-sensed forest cover data (KFS 2010; Kim et al. 2016);  
104 field data from the Korean National Forest Inventory (NFI) (National Institute of Forest  
105 Science, NIFoS 2011); topographic data; and data on restricted areas with impractical working  
106 conditions, protected areas, and areas of national interest for nature conservation (Kim et al.  
107 2016); we spatially integrated these data sets with a 0.01° grid size ( $\approx$ 1 km). We also here  
108 account more detailed forest specific information, such as the distribution of stand age, tree  
109 species, site indices, and management types, in both this assessment and earlier reports (Kim  
110 et al. 2019) using downloaded the data from the Integrated Forest Growth Information System  
111 (<http://map.forest.go.kr/>).

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### 113 **2.2. Climate data**

114 We also applied two climate data sets for assessing trade-offs and synergies among  
115 the ecosystem services. In the first data scenario, CC, climate conditions were assumed to  
116 repeat as current levels (2001–2010) by every ten years from 2011 to 2050 in South Korea,  
117 specifically, mean temperature and precipitation of 10.2 °C and 1,150 mm. The data for the

118 second climate scenario were Representative Concentration Pathway 8.5 (RCP8.5). In the *Fifth*  
119 *Assessment Report* of the Intergovernmental Panel on Climate Change (IPCC), RCP8.5  
120 assumes that greenhouse gases (GHGs) will continue to be emitted at the current rates and the  
121 carbon dioxide concentration will reach 940 ppm; in this scenario, Korea's mean temperature  
122 and precipitation changed from 10.2 °C and 1,150 mm in 2010 to 12.9 °C and 1,802 mm in  
123 2050, both respectively (Fig. S2). The Korea Meteorological Administration provided the high-  
124 resolution (1-km spatial resolution) raster-based monthly mean temperature and precipitation  
125 data (<https://data.kma.go.kr/>).

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### 127 **2.3. Forest management scenarios**

128 To assess changes among different ecosystem services, we defined three forest  
129 management scenarios classified by annual harvest based on South Korean' forest management  
130 policies. Under the first scenario, Sc1, an overprotective forest area is managed with controlled  
131 management. For this scenario, we assumed that the total forest area was the defined protected  
132 area without harvest during study periods, and therefore, we estimated annual IWP by harvest  
133 activities as zero under Sc1.

134 For the second scenario, Sc2 [=business as usual (BAU)], we held the current harvest  
135 area at 15,000 ha per year; in this scenario, the KFS maintained annual harvest areas at 2010s  
136 levels. In the third scenario, Sc3 (=adaptive management), the harvest area increased to 35,000  
137 ha per year based on the 6th Basic Forest Plan (KFS 2017); this scenario reflected harvest  
138 policies based on improved forest management strategies.

139 In this study, we divided the forest area into two types—manageable and restricted—  
140 based on previous studies (Kim et al. 2016). Based on forest legislation and geographic  
141 conditions, the forest area with harvest activities occupied approximately 3,138,000 ha,

142 approximately 51.7% of the total forest area (Fig. S1d). The KFS's sustainable forest  
143 management law authorizes clear-cut harvest at the final age of maturity for each tree species  
144 only in managed areas, and forest managers oversee each stand according to the cycle of final  
145 felling and regeneration. We did not include natural disturbances and changes in land use/cover  
146 such as afforestation or deforestation in the simulations in this study, assuming, rather, that total  
147 forest area remained constant over time.

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#### 149 **2.4. Quantifying ecosystem services**

150 Forest maturity and management activities affect forest-based ecosystem services  
151 (Bugmann et al. 2017; Fig. S3). For example, forest C stocks and recreation services increase  
152 with forest maturity (Norman et al. 2010), but Cse remains slow in the “stand imitation phase”,  
153 becomes faster beyond that, and shoots up in “stem-exclusion phase” to “mature phase” until  
154 it reaches a peak, after which Cse declines (Oliver et al. 1996); harvesting produces timber and  
155 thus decreases Cse in the near term, but more C will be absorbed from the atmosphere in the  
156 future (Kim et al. 2017). For this study, we assessed the changes in forest ecosystem services  
157 under each scenario based on the impacts of maturity and harvest management on the services  
158 (Fig. S3 and S4).

159 We used the integrated tool based on models and methods for spatially explicit  
160 simulation to assess the corresponding changes in forest ecosystem services. This tool  
161 comprises three main modules: dynamic growth model for calculation of storage and yield, a  
162 water yield model for estimation of freshwater supply, and recreation service model for  
163 assessment of potential recreation value (Fig. S5). We used the tool to simulate forest growth  
164 and management under combinations of our three different management and two climate  
165 scenarios (Sc1-CC, Sc1-R85, Sc2-CC, Sc2-R85, Sc3-CC, and Sc3-R85) and assessed the

166 impacts on the five ecosystem services. Specifically, we mapped, quantified, and compared the  
167 consequences for the ecosystem services under the different scenarios, which allowed us to  
168 compared the trade-offs and synergies among different ecosystem services and scenarios. We  
169 conducted all spatial analysis, buffer analysis, raster calculation, zonal statistics, and  
170 visualization of maps in ArcGIS 10.5 (ESRI 2015). A detailed description of the main models  
171 and all abbreviations for this study are presented in supplement materials (Table S1 and Fig.  
172 S4).

173

#### 174 **2.4.1. Simulating forest growth and management**

175 In this study, for application of dynamic growth model, we constructed the forest  
176 stands in 1 km × 1 km grids and described the state of each grid with four variables with varying  
177 values: forest type [seven major forest types: red pine (*Pinus densiflora*), Japanese larch (*Larix*  
178 *kaempferi*), Korean pine (*Pinus koraiensis*), cork oak (*Quercus variabilis*), Mongolian oak  
179 (*Quercus mongolica*), mixed forest A (Mixed-A: red pine and cork oak)], and mixed forest B  
180 (Mixed-B: red pine and Mongolian oak), site index, stand age, and management type using the  
181 fifth NFI and the fifth Forest Cover Map (FCM). In addition, each grid was spatially linked to  
182 climatic and topographic conditions from digital elevation model data and climate data. A  
183 detailed description of the processes of data preparation is given in Kim et al. (2019). We first  
184 assigned values for each of the four variables to each grid point. To simulate the future forest  
185 development, we also projected the grids' future states based on the forest management  
186 treatments specified for individual grid, i.e., final felling or no action. Beginning with each  
187 grid's initial status in the 2010 year, CS, Cse, and IWP of each grid were simulated considering  
188 climate and management scenarios annually until the 2050 year using the dynamic growth  
189 model. In addition, some of the outputs from this simulation used as the input data for assessing

190 WS and RE value through the integration tool.

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#### 192 **2.4.2. Forest biomass and C stocks**

193 To simulate stem volume and carbon stock annually under climate change and  
194 management scenarios, we integrated four independent Korea-specific models: diameter at  
195 breast height (DBH) growth, tree mortality, growth modifier, and height growth as developed  
196 earlier (Kim et al. 2019). This model simulates forest volume and C stocks as follows: (1)  
197 estimated mean tree height growth based on stand age and site index of each stand using  
198 regression models, as suggested by the KFS (2009); (2) estimated stand density (trees per ha)  
199 change using a mortality model (Kim et al. 2017); (3) estimated stand mean DBH growth  
200 reflecting topographic and annual climatic conditions using regression models and a growth  
201 modifier developed by Piao et al. (2018) and Kim et al. (2019); and (4) estimated forest volume  
202 and carbon stocks using the outputs from the previous three steps, regression models, basic  
203 wood density, and biomass extension factors developed by NIFoS (2014; Table S2). Kim et al.  
204 (2019) described and validated the performances of this model, and Kim et al. (2019) give a  
205 detailed description of the integrated model.

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#### 207 **2.4.3. Yield and IWP production**

208 In this study, we considered only differences in rotation time for each stand according  
209 to tree species based on the legal final cutting age (Table S3). The state variable  $x$  represents  
210 the stand age on a 100-ha stand by each grid size in this study; there are 201 age classes and  
211 each age class represents one age, spanning tree ages from 0 to 200. The decision variable  $d$  is  
212 set to 0 for keeping the stand or 1 for fell the stand with the decision made at the start of the  
213 stage knowing the current age of trees in the stand; tree age is zero at the start of a stage if trees

214 are felled at that time. We implemented a clear-cut harvest at the final age of maturity for each  
 215 tree species (Table S3), and based on the management scenarios assumed in Sc2 and Sc3, the  
 216 maximum annual harvest areas for the scenarios were, respectively, 15,000 ha and 35,000 ha.  
 217 In the study, if more forest stands than the maximum reached final cutting age, we applied  
 218 harvesting to the top 150 or 350 grids with the highest stand age and volume stocks by  
 219 management scenario. The simulations in this study did not include any biomass loss in  
 220 harvesting.

221

$$222 \quad H_t = \sum f(\text{selection } x_{it}, x_{it} \geq FC), H_t \leq H_{max} \quad (1)$$

$$223 \quad IWP_t = \sum f(\text{biomass } x_{it}, x_{it} \in H_t) \quad (2)$$

224

225 where  $i$  is the identification number of a grid in this study;  $f(\text{selection})$  is a logical function as  
 226 the selection of harvest area which are the top 150 or 350 grids with the highest stand age and  
 227 volume stocks among  $x_i$  in year  $t$ ;  $H_t$  is the harvest area in year  $t$ ;  $x_{it}$  is the stand age of grid  $i$  in  
 228 year  $t$ ;  $FC$  is the legal final cutting age of each tree species (Table S3);  $H_{max}$  is the maximum  
 229 of annual harvest area for each scenario;  $f(\text{biomass})$  is a logical function as return the stem  
 230 biomass of  $x_i$  in year  $t$ ; and  $IWP_t$  is the industrial wood production in year  $t$ .

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#### 232 **2.4.4. Estimation of forest recreation service**

233 Many researchers have linked forest maturity, accessibility, and diversity of structure  
 234 with high RE value (Hörnsten and Fredman 2000; Nielsen et al. 2007; Zandersen and Tol 2009).  
 235 In this study, we quantified RE of each grid annually based on functions using geographic and  
 236 environmental factors (Eq. 2 and 3; Table S4). Firstly, we consider the recreational  
 237 opportunities and accessibility to the forests to assess the potential RE based on a previous

238 study (Pang et al. 2017). The grids within 2 km of housing area or within 300 m along each  
 239 side of small roads are only considered to evaluate RE in this study. Secondly, we used the  
 240 stand age and volume for considering the diversity of forest maturity. We determined that  
 241 forests more than 70 years old or higher stand volume than the reference stand volume were  
 242 valuable RE. The stand volume of applicable forest is divided by the reference stand volume  
 243 in the assessing process of RE in order that forest quality should be reflected (Eq. 2). The  
 244 reference stand volume for each forest type is calculated that the mean stand volume for the  
 245 NFI plots that stand age is older than 70 stand age minus their one standard deviation (Table  
 246 S5). In addition, to reflect the different preferences of visitors for forest types such as  
 247 coniferous, broad-leaved, and mixed forest,  $R_p$  is multiplied by 0.75, 1, 1.25 for these forest  
 248 types (Eq. 3). These values were computed based on the difference of willingness to pay from  
 249 the analysis in the economic valuation of the recreational benefits (Nielsen et al. 2007).  
 250 Although in that research, there is no specific value for the coniferous forest, we estimated the  
 251 relative value for coniferous forest using the same ratio between the mixed forests and  
 252 monoculture of broadleaves. Thirdly, we evaluated the visual diversity and spatial diversity of  
 253 each grid in the assessment of RE. Each grid was computed a diversity index for the values  
 254 within a specified neighborhood (3×3) for each input cell position. The grid with one, two, and  
 255 three forest types in the neighborhood grids were valued as 0.75, 1.0, and 1.25, respectively.  
 256 We used land cover data on residential areas and road systems together with stand age data  
 257 from the forest cover map in the 5th NFI and the output from the dynamic forest growth model  
 258 to identify potential recreation area. Based on these considerations, the RE value for each grid  
 259 was computed as follow (Eq. 3):

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$$R_{it} = f(\text{Age}_{it} \text{ or } \text{Volume}_{it})/F_r \times f(\text{Accessibility}_i) \times T_i \times N_{it} \quad (3)$$

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where  $f(\text{Age}_i \text{ or Volume}_i)$  is a logical function as if stand age or stand volume ( $\text{m}^3 \text{ ha}^{-1}$ ) of  $i$  grid in year  $t$  is older than 70 years or higher volume than reference stand volume ( $\text{m}^3 \text{ ha}^{-1}$ ), stand volume of  $i$  grid in year  $t$  is returned, if it is not, the zero value is returned;  $F_r$  is reference stand volume for forest type;  $f(\text{Accessibility})$  is a logical function that if  $i$  grid is within a 2 km radius of housing areas or linked to the areas within 300 m radius from a road network, the one value is returned, if it is not, the zero value is returned;  $T_i$  is preferences for forest type of  $i$  grid;  $N_{it}$  is a diversity index for  $i$  grid in year  $t$ ; and  $R$  is estimated recreation service value. We considered that forest area, residential area, and road network were constant in the analysis.

#### 2.4.5. Estimation of fresh water supply

We calculated the WS based on the annual water yield after multiplying the above- and underground run-off coefficient by the annual precipitation in the spring season (March–May). To reflect age-dependence and management influences on WS from the forest, we applied the mathematical model described by Tian et al. (2008). They found that water distribution on forest floors in differently aged forest plantations was mainly influenced by human cultivation activities in young stands and by stand structure in mature stands. Thus, this method was well suited for modeling forest management activities such as harvest and reforestation, adapted to South Korea in this case, using the evapotranspiration (ET) coefficient of major tree species. However, it had a limitation that the research was only focused on one tree species as Chinese fir. Therefore, we applied the modifier function was developed by Kim (2018) to expand spatially from the plot-level methodology to the national level (Eq. 4). The freshwater supply model involves many factors, such as forest type, stand age, topography, monthly precipitation, and coefficients for estimation evapotranspiration. Tian et al. (2008) and

286 Kim (2018) give a detailed description of the analysis processes and coefficients for Eq. 4 and  
287 5.

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$$289 \quad MD_i = \frac{1-ET_i}{1-ET_{CF}} \cdot \frac{C_i}{C_{25}} \quad (4)$$

$$290 \quad WS_t = \sum P_{it} \cdot (b_{ijt0} + b_{ijt1} \cdot P + b_{ijt2} \cdot P^2 + \dots + b_{ijtn} \cdot P^n) \cdot MD_i \quad (5)$$

291 where  $MD_i$  is the modifier at grid  $I$ ;  $ET_{CF}$  is the evapotranspiration (ET) coefficient of Chinese  
292 fir (0.77);  $ET_k$  is the  $ET$  coefficient of each Korean major tree species (0.70, 0.77, 0.82, 0.60,  
293 and 0.69 for red pine, Japanese larch, Korean pine, oak, and mixed forest, respectively) at grid  
294  $i$ ;  $C_i$  is the runoff coefficient according to the slope at grid  $i$ ;  $b_n$  are the coefficients for grid  $i$   
295 with stand age  $j$  in year  $t$  (Tian et al. 2008);  $n$  represents the degree of a polynomial function;  
296  $P$  is the precipitation at  $i$  grid during spring season in year  $t$ ; and  $WS_t$  is the freshwater supply  
297 in  $t$  year.

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### 299 **3. Results and Discussion**

#### 300 **3.1. Spatiotemporal distribution of ecosystem services**

##### 301 **3.1.1. C storage**

302 We estimated CS changes by forest factors such as forest rotation age and by climate  
303 change scenario (Fig. 1a). Total forest CS under Sc1-CC and Sc1-R85 increased, respectively,  
304 from 400.0 Tg C to 814.1 Tg C and 780.9 Tg C at the rates of 10.35 and 9.52 Tg C yr<sup>-1</sup> during  
305 2010–2050. Our results projected relatively high CS increases between the western and eastern  
306 parts of South Korea but not in the mountainous area (Fig. S6); one of the main reasons for  
307 these results is that the forests in this area are younger overall than the mountain forests (Kim  
308 et al. 2019). We tried to validate the model separately in this study by comparing estimated CS

309 with the Statistical Yearbook of Forestry (KFS 2019). The CS for all South Korean forests  
310 increased from 405.7 to 493.8 Tg C between 2010 and 2017, and in simulation Sc2-CC, CS  
311 increased from 400.0 to 491.3 Tg C during that period. Therefore, it could be inferred that the  
312 model successfully reflected the change in volume and C trends of Korean forests by time  
313 series, although uncertainties remain for individual stands spatial consistency between  
314 observed data and model results.

315 The differences in the C stocks among the studied management scenarios are depicted  
316 in Fig. 1a and Fig. S7. Estimated CS in 2050 under scenarios Sc2-CC and Sc3-CC was 746.6  
317 and 659.9 Tg C, respectively. Typically, CS varies based on growth and harvest activities  
318 (cutting by final maturity age) and considering climate change in Sc2-R85 and Sc3-R85, the  
319 projected CS values for 2050 were 718.6 and 648.1 Tg C, respectively. The estimated  
320 differences between their C stocks in 2050 shown by climate change were  $-27.9$  and  $-11.8$  Tg  
321 C, respectively. Our results showed that management activities might help lessen the negative  
322 impacts of climate change on forest growth and C stocks. Based on observation data, climate  
323 change impacts on tree growth of major temperate forests in South Korea are different  
324 depended on tree species and geographical conditions (Kim et al. 2019, 2020). Therefore,  
325 decision-makers in forest management have to need the insight to maximize the positive effects  
326 of climate change on the forest with a long-term perspective (Lindner et al. 2014).

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### 328 **3.1.2. C sequestration**

329 As shown in Fig. 1b, the annual Cse in South Korean forests tended to decrease  
330 gradually from 2010 to 2050, unlike the results for CS; this is because the annual forest growth  
331 and Cse were related to stand age in our approach. In most cases, the age response of a C sink  
332 to a new steady state consists of a rapid initial increase followed by a gradual decline (Hudiburg

333 et al. 2009). Changes in annual Cse are also affected by forest management activities. Within  
334 the three management scenarios without climate change, i.e., Sc1-CC, Sc2-CC, and Sc3-CC,  
335 the annual Cse tended to increase in the presence of forest management activities (Fig. 1b), and  
336 without these activities, decreased from 11.58 Tg C yr<sup>-1</sup> in 2010 to 5.74, 6.28 and 6.80 Tg C yr<sup>-1</sup>  
337 in 2050, respectively, during the study period. In our simulation, the harvested forest area in  
338 *i* year would be reforested automatically with one-year-old stand age in *i+1* year. Therefore,  
339 the over matured forests with low Cse capacity could be changed to young forests with high  
340 Cse capacity by harvest management. These results indicate that the harvest activity could help  
341 improve the Cse and imbalanced age structure of Korean forests in the long-term perspective.  
342 Approximately 67% of Korean forests had trees between 40 and 50 years old (KFS 2016) in  
343 the 2015 year and thus will require infrastructure and systems for larger future harvesting areas  
344 than after the 2010s, as provided by the forest management practice guide in South Korea.

345

### 346 **3.1.3. Freshwater supply**

347 The estimates of spring WS in the next four decades clearly showed gradual decreases  
348 over time (Fig. 1c). In scenario Sc1-CC, the WS was 4.53 billion m<sup>3</sup> year<sup>-1</sup> in 2010 and 4.15  
349 billion m<sup>3</sup> year<sup>-1</sup> in 2050; this decline is linked to changes in forest structure over time  
350 considered in the simulation of water supply model. In the water supply model used in this  
351 study, the WS to precipitation ratio (%) in a forest area decreased steadily with forest aging,  
352 and other researchers have shown obvious declines in water supply as forests aged and cover  
353 and canopy increased (Cademus et al. 2014). The reduced water yields following canopy  
354 expansion derive from the changes in the balance between rainfall, evaporation, and system  
355 runoff response (Ilstedt et al. 2016); higher evapotranspiration, reduced water availability in  
356 soils and groundwater, and changes in energy fluxes (e.g., albedo, sensible, and latent heat) are

357 some of the main mechanisms driving these changes (Ellison et al. 2017). In addition, these  
358 changes can be explained by the substantial decrease in rainfall amount and intensity that  
359 reaches the forest floor as the stand structure develops.

360         However, WS in scenarios Sc2-CC and Sc3-CC were projected to start to increasing  
361 gradually before stabilizing in accumulated harvested areas during the period from the 2030s  
362 to the 2040s. The simulation results showed that the 2050 WS estimates under Sc2-CC and  
363 Sc3-CC were 4.32 and 4.50 billion m<sup>3</sup> year<sup>-1</sup>, respectively, accounting for 95.3% and 99.4% of  
364 the freshwater supplies in 2010. These results are closely linked to ecological mechanisms that  
365 the changes in forest structures caused by aging or disturbances could influence on runoff  
366 characteristics in forests (Ellison et al. 2017; Tian et al. 2008). In addition, based on several  
367 articles from the past decade, impact of afforestation or reforestation on water yield is negative  
368 such as reduction of downstream water availability, while forest clearing results in increased  
369 streamflow (Farley et al. 2005). Our results suggest that forest regeneration might help improve  
370 the national water cycle and reduce the damage from spring droughts.

371         Although the forest WS estimates of this study showed a broadly similar pattern with  
372 previous researches (Jung et al. 2009; Kim et al. 2010; Kim et al. 2017), there are some  
373 differences in the results of each study. There were two possible reasons for this. As shown in  
374 Table S6, each study used different models and methodologies to estimate forest WS. This  
375 probably contributes to making a difference in the estimated forest WS. The other possible  
376 reason was a difference in the research year and periods. As noted earlier, the amount of forest  
377 WS is greatly influenced by precipitation patterns. Therefore, our results for WS could be  
378 considered reliable even though it is based on indirect validation.

379

#### 380         **3.1.4. Industrial wood production**

381 IWP differences in each scenario are illustrated in Fig. 1d; the results show that yields  
382 were higher in Sc3 than in Sc2. The annual mean IWP under Sc2-CC increased from 2.46  
383 million m<sup>3</sup> in the 2010s to 3.36 million m<sup>3</sup> in the 2040s, and in comparison, the annual mean  
384 IWP under Sc3-CC in the 2040s was 7.67 million m<sup>3</sup>, 228.2% higher than that under Sc2-CC.  
385 This was because the annual harvest area in Sc3 was larger than that in Sc2 in our assumption.  
386 According to the national forestry report, the total IWP from South Korean forests increased  
387 from 1.14 to 2.29 million m<sup>3</sup> between 2010 and 2017 (KFS 2019), and in simulation Sc2-CC,  
388 it increased from 0.67 to 3.04 million m<sup>3</sup> during that period. There are two possible reasons for  
389 these differences. First, they could be attributable to our method for selecting the target harvest  
390 areas; in our simulations, harvest management was prioritized based on stand age and stem  
391 volume, and therefore, the mean stem volume of forest areas selected for harvest in the model  
392 could be higher than that of actual harvest areas. The other possible reason could be related to  
393 the biomass loss rate in actual harvest activity. In our simulation, we did not consider biomass  
394 loss in harvest management.

395

### 396 **3.1.5. Forest recreation**

397 We observed the total RE value in every scenario tended to increase gradually from  
398 2010 to 2050. The potential recreation area and total RE value in Sc1-CC were estimated to be  
399 689,400 ha and 0.75 million in 2010, and it increased to 3,431,100 ha and 5.15 million in 2050.  
400 Our estimated RE value in the 2010s was very low because stand age and volume are  
401 considered one of the main factors in assessing RE. Korea's forests have spatial heterogeneity  
402 in stand age caused by profound historical deforestation before the 1960s and national  
403 plantation programs after the 1970s (Kim et al. 2019). In the NFI data, the accessible lowland  
404 forests are on average 10 years younger than mountainous forests that are distant from main

405 residential areas (NIFoS 2011). As shown in Fig. 1e, changes in RE are also affected by forest  
406 management activities. The simulation results showed that estimated RE value under Sc2-CC  
407 and Sc3-CC in 2050 were 4.95 and 4.18 million, respectively, accounting for 93.3% and 81.2%  
408 of the RE under Sc1-CC in 2050. Owing to the larger areas of final felling and, thus of younger,  
409 less mature trees, the RE value in Sc3 was much smaller than that in Sc2 (Fig. 1e).

410         Difference patterns between the climate scenarios in the estimated RE value are shown  
411 (Fig. 1e). In simulation periods, the overall results show that the total RE value under Sc2 and  
412 Sc3 were higher in CC than in R85. However, the total RE value in Sc1-CC is estimated as  
413 slightly lower than in Sc1-R85. This was because each grid of forest area is given different  
414 weight depending on stand volume, dominant tree species, and diversity of species composition  
415 with the neighbor area (diversity index in Eq. 3) in the process of estimation RE value. In our  
416 simulation results, the overall growth of broad-leaved forests with higher weight values for RE  
417 than coniferous forests were positive effected by climate change according to R85 (Fig. 1e). In  
418 addition, the overall diversity index was estimated to increase gradually from 2010 to 2050 due  
419 to without harvest activity in Sc1-R85. In Sc2-R85 and Sc3-R85, the climate change impacts  
420 on forests were the same as in Sc3-R85, but the diversity index and potential recreation areas  
421 were lower, due to harvest management than in Sc1-R85. Consequently, these negative effects  
422 on RE from the harvest in Sc2-R85 and Sc3-R85 were estimated higher than the positive impact  
423 of climate change.

424

### 425 **3.2. Variations in trade-offs and synergies**

426         We compared the projections for the five ecosystem services in the 2040s with the  
427 situation in the 2010s (2010–2017), the base period for assessment, and the differences are  
428 shown in Fig. S6. We identified the trade-offs and synergies among multiple services by forest

429 aging by comparing the conditions in 2015 with those in the 2040s under Sc1-CC; as expected,  
430 we observed clear synergy between CS and RE in this study. The scenarios further showed  
431 trade-offs between CS and RE and IWP on one side and between Cse, IWP, and, to some extent,  
432 WS on the other. In the 2040s under Sc1-CC, Cse decreased by approximately 43% compared  
433 with the base year, thus negatively affecting South Korea's achieving its nationally determined  
434 contributions. South Korea's NDCs include the target of reducing GHG emissions, excluding  
435 land use, land-use change, and forestry, by 37% below the business-as-usual emission (850.6  
436 MtCO<sub>2</sub>eq) by 2030.

437 Fig. S6 presents the temporal variations in the trade-offs and synergies among multiple  
438 ecosystem services by climate change from the 2020s to the 2040s. WS correlated strongly  
439 with precipitation levels. For instance, in the RCP8.5 scenarios, spring precipitation in South  
440 Korea was considerably higher in the 2020s than in the 2010s, but projected WS decreased to  
441 nearly the same level as for the 2010s during the period from the 2020s to the 2040s. Other  
442 ecosystem services except RE were estimated to decrease gradually over time because of  
443 climate change. Estimated average CS, Cse, and IWP under Sc2-R85 in the 2040s were 30.7  
444 Tg C, 6.51 Tg C yr<sup>-1</sup> and 3.11 million m<sup>3</sup>, and they decreased by 3.3%, 2.7%, and 7.3%  
445 compared with Sc2-CM in the 2040s, respectively.

446 In Sc2-CC and Sc3-CC in the 2040s, Cse improved by approximately 6.6% and 24.0%  
447 and WS by approximately 4.0% and 8.4%, respectively, compared with in Sc1-CC in the same  
448 decade. In addition, the ratio of CS, Cse, WS, IWP, and RE of Sc3-to-Sc2 in the 2040s was  
449 estimated as 88.4%, 114.7%, 105.4%, 228.2%, and 86.5%, respectively. These results showed  
450 the intensified harvest activity might provide a negative effect on Cse in the short term, but the  
451 long-term effect could be positive. These results also implied that sustainable forest harvesting  
452 could help in maintaining ecosystem services such as Cse and WS.

453 In this research, we tried to handle trade-offs and synergies among five forest  
454 ecosystem services by aging, management, and climate change using various models. This is  
455 necessary for the integrated assessment of change in forest ecosystem services and providing a  
456 revealing insight into the policymakers. In addition, although this research focused on Korean  
457 forests, the results of the changing trend among forest ecosystem services are similar to other  
458 previous research for temperate forests (Bugmann et al. 2017; Pang et al. 2017). Therefore, our  
459 results cannot be only helpful for Korean forests but also temperate forests in the world.  
460 However, this strong point of our research could lead to a bit of complicated to describe the  
461 interactions between the different ecosystem services. Therefore, this limitation should be  
462 solved and improved in further research.

463

### 464 **3.3. Implications for forest management**

465 In order to measure the function of a forest as a carbon sink, we used biomass expansion  
466 factors to estimate the carbon content of a whole tree using stem volume as input data. Thus,  
467 in this study, we assessed only CS and Cse in the tree layer; the total forest carbon stock would  
468 have been considerably higher if we had included soil carbon, litter carbon, and ground  
469 vegetation (Kim et al. 2016). Moreover, considering carbon stored in wood products and  
470 substitution effects (such as forest regeneration) from using forest biomass as raw material  
471 instead of fossil fuels would have affected the carbon balance (Kallio et al. 2013).

472 This study simulated RE that were driven mainly by the forest's structures. In case of  
473 the RE, the provisioning potential could rely on forest's structure resulting from the dynamics  
474 by climate change, but the consumer's perceptions of ecosystem services can be enlarged by  
475 various climate resilience management and policies such as application of forest therapy,  
476 increasing forest education, investment of forest village, and forest tourism management

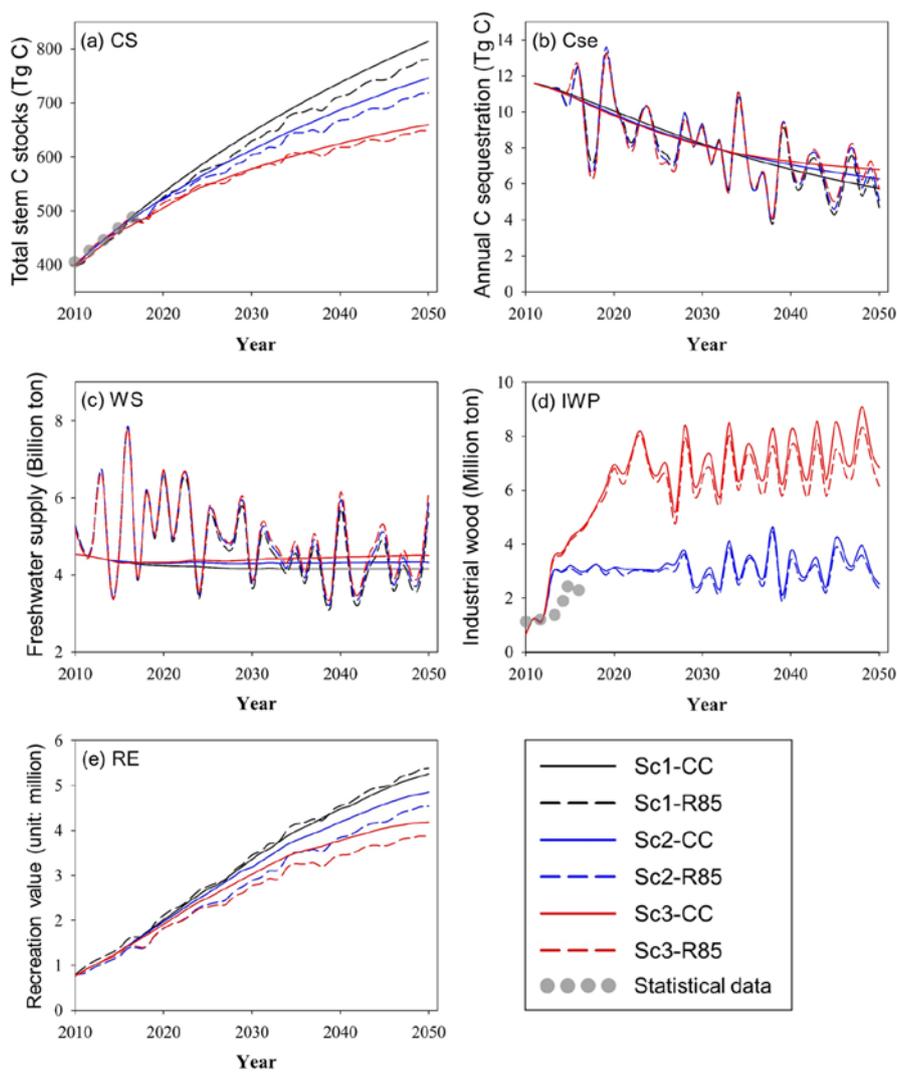
477 (Nielsen et al. 2007). Therefore, further research on the recreational values of Korean forests  
478 associated with identifying and assessing public preferences is suggested.

479         In this study, we assessed only IWP from stem volume, but final felling also produces  
480 harvest residues, i.e., tops, branches, and foliage. These have diverse uses such as for producing  
481 bioenergy, bio-compost, sawdust, and wood chips. Forest bioenergy is generally considered  
482 renewable, carbon-neutral energy, which means that forest bioenergy could mitigate climate  
483 change due to the neutrality of GHG emissions in the whole harvest and regeneration cycle.  
484 Our results indicate that given the abundant forest cover in South Korea, there is a substantial  
485 potential for bioenergy production, which could not only reduce fossil fuel emissions but also  
486 contribute to meeting any targets agreed to under potential climate change mitigation  
487 agreements.

488         However, increased extraction of harvest residues can be expected to have  
489 environmental impacts. For example, these could affect soil acidification, causing the decline  
490 of forest productivity (de Jong et al. 2017). In addition, residues contribute to the supply of  
491 deadwood, an important forest biodiversity component (Sullivan et al. 2011). In addition,  
492 increased surface fuel and a homogenized forest structure resulting from management practices  
493 such as timber harvest could lead to the accumulation of forest fuels that are playing a major  
494 role in increasing wildfire size and severity in many semiarid forest types (Allen et al. 2002).  
495 Therefore, future harvest policies should consider the balance between the extraction targets,  
496 such as commercial targets to produce IWP and fuel reduction targets to reduce wildfire hazards,  
497 and residue retention targets to maintain habitat and other forest functions.

498         Although we focused on only one type of forest management in this study, it is  
499 necessary to consider the correlation of other management practices and changes in ecosystem  
500 services additionally for improving or maintaining the resilience of the forest ecosystem.

501 Timber production by thinning and reforestation with trees of different species could help to  
 502 lessen the negative effects of biomass production by improving the soil biota diversity  
 503 (Vicente-Vicente et al. 2019). In addition, the techniques and systems for forest management  
 504 are also important to maintain timber production while minimizing the negative effects on  
 505 ecosystem services. According to previous research, improved timber harvesting could have  
 506 relatively benefits to reduce logging emissions (Ellis et al. 2019), biodiversity (Bicknell et al.  
 507 2014), and soil erosion (Vicente-Vicente et al. 2019). Therefore, management-types and -  
 508 related techniques should comprehensively consider in further study or actual actions for  
 509 assessment of change in forest ecosystem services.



510

511 **Fig. 1.** Supply of ecosystem services such as (a) CS, (b) Cse, (c) WS, (d) IWP, and (e) RE  
512 during 2010–2050 in the forest management and climate change scenarios [CS: carbon stock,  
513 Cse: annual C sequestration, WS: freshwater supply, IWP: industrial wood production, and RE:  
514 potential recreation value].

515

#### 516 **4. Conclusion**

517 This study projected the changes among ecosystem services with related forest  
518 management strategies and climate change using a spatially explicit tool for applying a forest-  
519 centered approach. The results showed that different harvest policies and climate change had  
520 different effects on potential biomass extraction and other ecosystem services, and these effects  
521 changed over time. The forest management scenarios showed trade-offs between RE and CS  
522 on one side and between IWP, WS and, to some extent, Cse on the other side. In addition, this  
523 study showed that future climate change under RCP8.5 until 2050 in South Korea could have  
524 a significant negative influence on forest carbon uptake and storage and could offset the  
525 positive effects of harvest activity.

526 The scenarios could be adjusted and re-run to find more sustainable solutions such as  
527 forest management strategies for the spatial distribution of management activities for the  
528 equilibrium of demand and supply in ecosystem services. Our findings and tools could be  
529 helpful that the synergies and trade-offs between forest ecosystem services can be localized,  
530 quantified, and assessed for different land use-related policies and plans, thus providing  
531 decision support for sustainable forest management.

532

#### 533 **Acknowledgments**

534 This research is supported by the Korea Ministry of Environment under the “Climate

535 Change Correspondence Program (Project Number: 2014001310008)". This work was also  
536 supported under the framework of international cooperation program managed by the National  
537 Research Foundation of Korea (Project No. 2019K2A9A1A02094860).

538

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656 **Supplement materials**

657 **Table S1.** Description of the main models for integrated analysis in this study

Model	Description	Reference
Dynamic growth model	<ul style="list-style-type: none"> <li>• Contribution: Application to predict forest carbon stock and sequestration under management and climate change scenarios</li> <li>• Input: Environmental factors (Stand age, Site index, Tree species), Climatic factors (Temperature, precipitation), Management factors (Clear cut)</li> <li>• Output: CS, Cse, IWP</li> </ul>	Kim et al. 2019; Kim et al. 2017; Piao et al. 2019; NIFoS 2014
Water yield model	<ul style="list-style-type: none"> <li>• Contribution: Application and spatial expansion, from plot- to national-level, of developed model using spatial analysis and dynamic growth model</li> <li>• Input: Stand age, Tree species, Slope, Precipitation</li> <li>• Output: WS</li> </ul>	Tian et al. 2008; Kim 2018
Forest recreation index model	<ul style="list-style-type: none"> <li>• Contribution: Development of spatial assessment model for forest recreation index using ArcGIS</li> <li>• Input: Stand age, CS, Tree species, Site suitability, Tree species composition</li> <li>• Output: Recreation index</li> </ul>	Nielsen et al. 2007 Pang et al. 2017

658

659 **Table S2.** Parameters for stand volume and carbon storage (NIFoS, 2014)

Tree species	Coefficients			<sup>a</sup> BWD	<sup>b</sup> BEF	<sup>c</sup> R-S
	a	b	c			
Red pine	0.034	1.734	1.025	0.472	1.413	0.254
Japanese larch	0.005	2.458	0.904	0.453	1.335	0.291
Korean pine	0.046	1.732	0.896	0.408	1.812	0.283
Cork oak	0.053	1.810	0.881	0.721	1.338	0.324
Mongolian oak	0.098	1.406	1.135	0.663	1.603	0.388

660 <sup>a</sup>BWD: Basic wood density, <sup>b</sup>BEF: Biomass expansion factor, <sup>c</sup>R-S: Root-shoot ratio

661

662

663 **Table S3.** The final cutting age of major Korean tree species (KFS, 2015)

Tree species	Final age
Red pine	60
Korean pine	60
Japanese larch	50
Cork oak	60
Mongolian oak	60
Mixed-A	60
Mixed-B	60

664

665

666 **Table S4.** Data and parameters for assessment of recreational value in forests

Factors		Measurement	Reference
Geographic factors	Accessibility to the forest	The forests within a 2 km radius of housing areas or within 300 m along each side of a road network	Hörnsten and Fredman (2000)
Environmental factors	Maturity	Stand age for coniferous, broad-leaved, and mixed forest is older than 70 years old or stand volume is higher than specific value for each forest type	Pang et al. (2007)
	Species	The coniferous, broad-leaved, and mixed forest was evaluated that its potential RS was multiplied by 0.75, 1.0, and 1.25, respectively	Nielsen et al. (2007)
	Composition with the neighboring areas	The visual and structural diversity is evaluated based on a variety of forest types in the neighborhood area.	Nielsen et al. (2007), Zandersen and Tol (2009)

667

668

669 **Table S5.** The reference stand volume of forest types for assessment of recreation service in  
 670 South Korean forests (KFS, 2015)

Forest type	Reference stand volume (m <sup>3</sup> ha <sup>-1</sup> )	Mean volume (m <sup>3</sup> ha <sup>-1</sup> ) (std. dev.)
Coniferous forests	141.87	234.25 (92.38)
Broad-leaved forests	133.53	201.29 (67.76)
Mixed forests	145.38	214.87 (69.49)

671

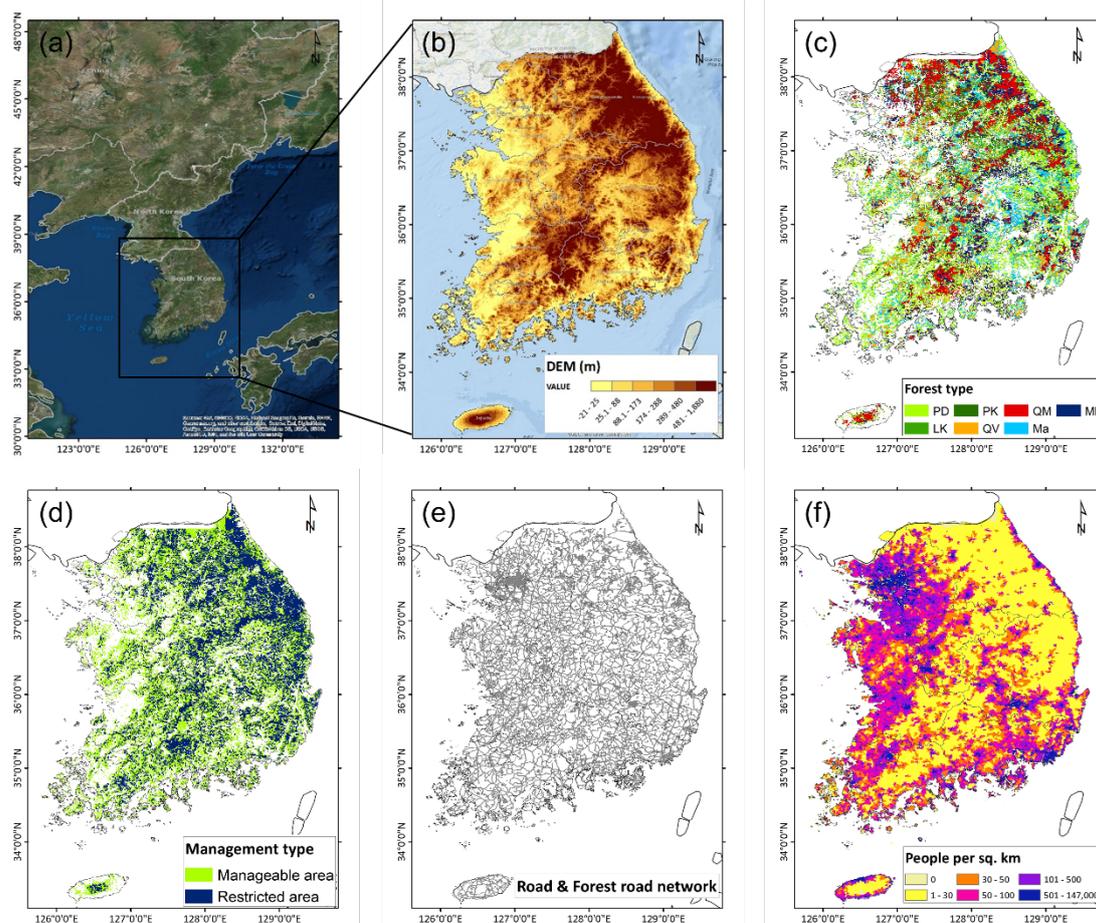
672

673 **Table S6.** Comparison of water yield from South Korean forests with that of previous studies

Year or period	Model	Estimate WY (yr <sup>-1</sup> )	Reference
2000s	InVEST	22,210 million m <sup>3</sup>	Kim et al. 2017
2011	Statistical model	19,060 million m <sup>3</sup>	Kim et al. 2010
2009	Statistical model	18,000 million m <sup>3</sup>	Jung et al. 2009
2011	This study	18,562 million m <sup>3</sup>	This study
2010s		20,562 million m <sup>3</sup>	

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677 **Fig. S1.** (a) Geographic location, (b) digital elevation model (DEM), (c) forest type distribution

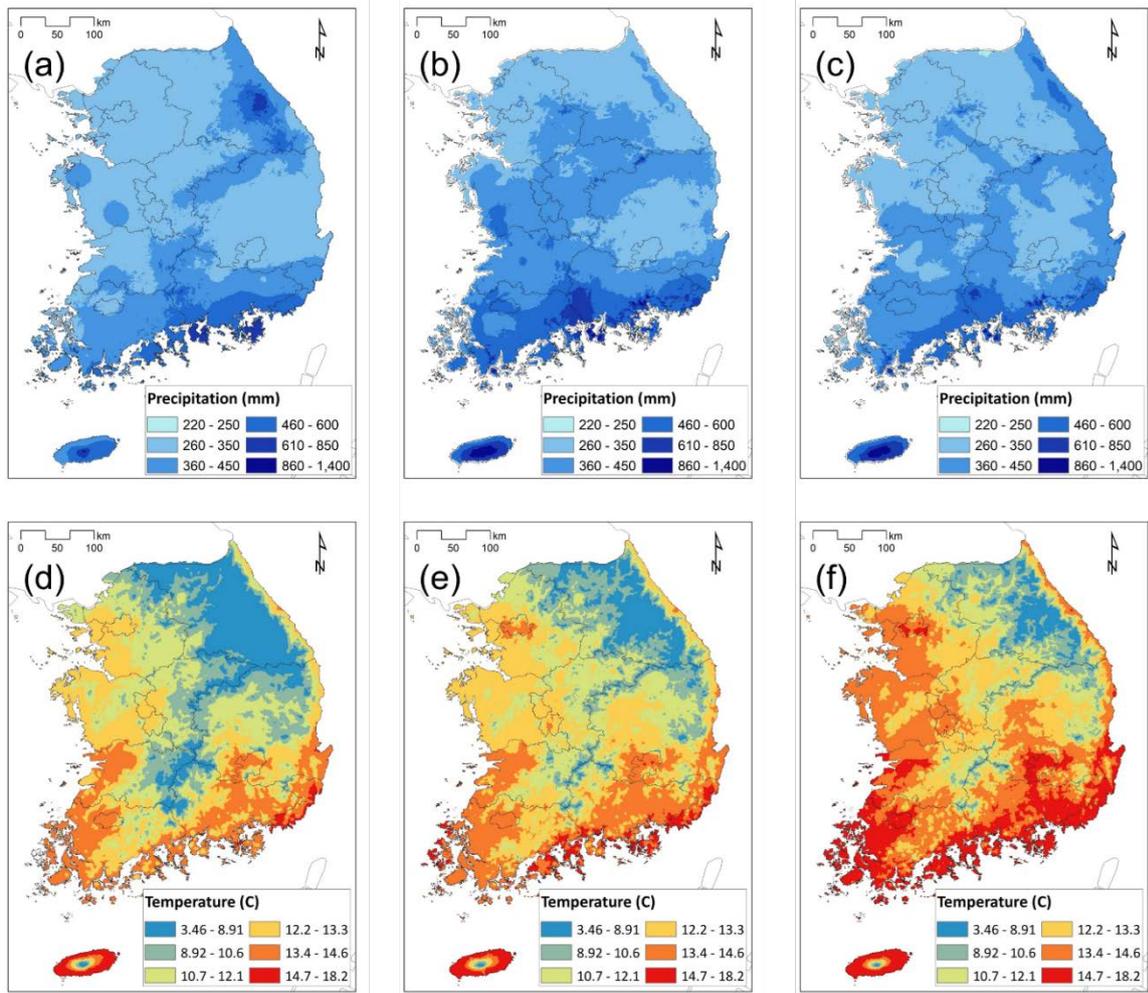
678 [(PD): *Pinus densiflora*; (PK): *Pinus koraiensis*; (LK): *Larix kaempferi*; (QM): *Quercus*

679 *mongolica*; (QV): *Quercus variabilis*); (Ma): Mixed forest (PD and QV); (Mb): Mixed forest

680 (PD and QM)], and (d) management type distribution in South Korea. Map of (e) road and

681 forest road system, and (f) population density.

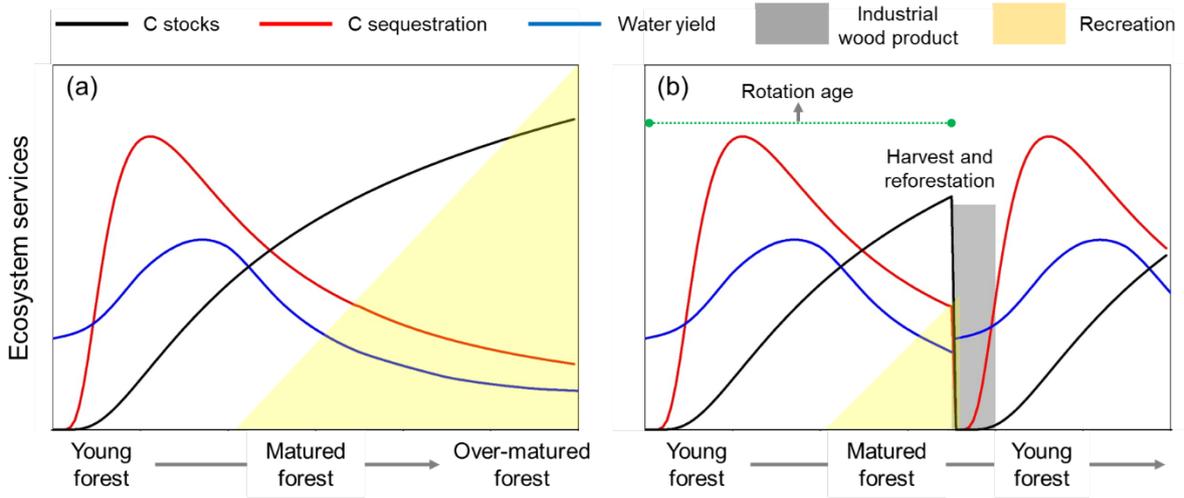
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683

684 **Fig. S2.** The projection of mean spring (March-May) precipitation and temperature over South  
 685 Korea in the 2000s (a,d), 2020s (b,e), and 2040s (c,f) under Representative Concentration  
 686 Pathway 8.5 (RCP8.5)

687

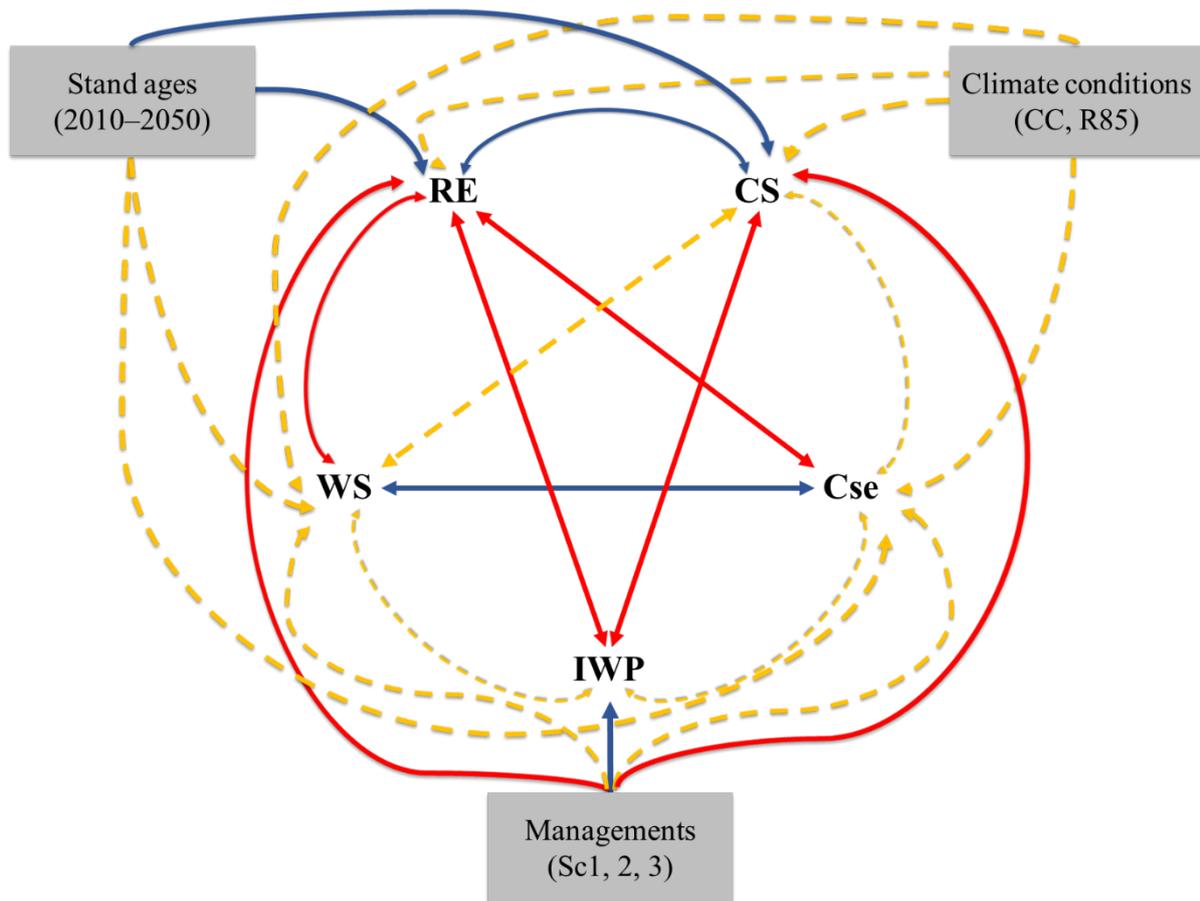


688

689 **Fig. S3.** Forest ecosystem services of (a) a non-harvest forest and (b) a managed forest

690 relative to stand maturity.

691



692

693 **Fig. S4.** Flow chart of management, age, and climate change effects on forest ecosystem

694 services. Blue and red solid arrows indicate positive and negative effect. Orange dashed arrow

695 means an intermediate effect. Abbreviations: CS – carbon storage, Cse – carbon sequestration,

696 IWP – industrial wood production, WS – fresh water supply, RE – forest recreation, Sc1-3 –

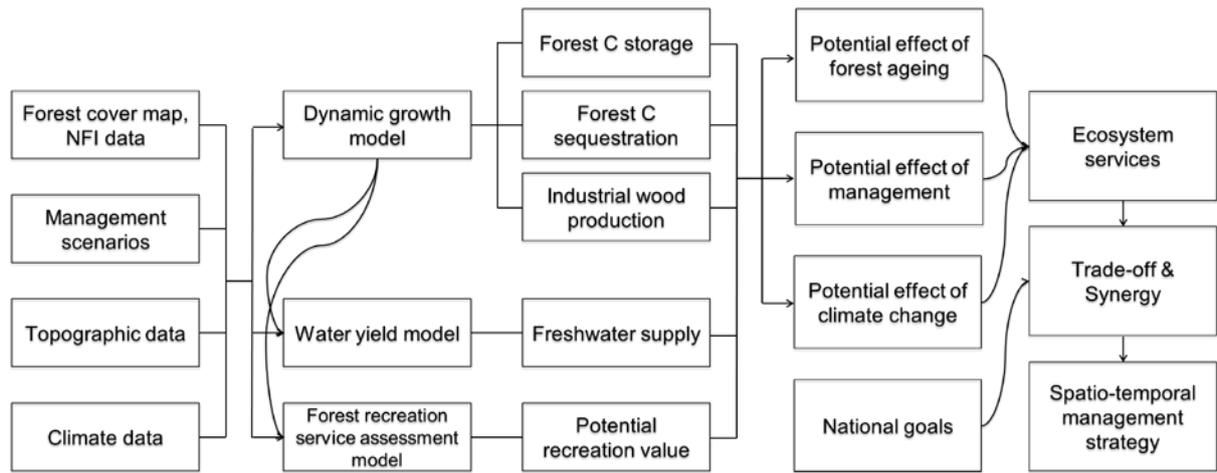
697 these are management scenarios that the Sc1, Sc2 and Sc3 are defined by the maximum annual

698 harvest area (Sc1: zero, Sc2: 15,000 ha, Sc3: 35,000 ha), CC – climate conditions remained at

699 current levels (2000–2010) until 2050, R85 – the projected climate conditions by

700 Representative Concentration Pathway 8.5

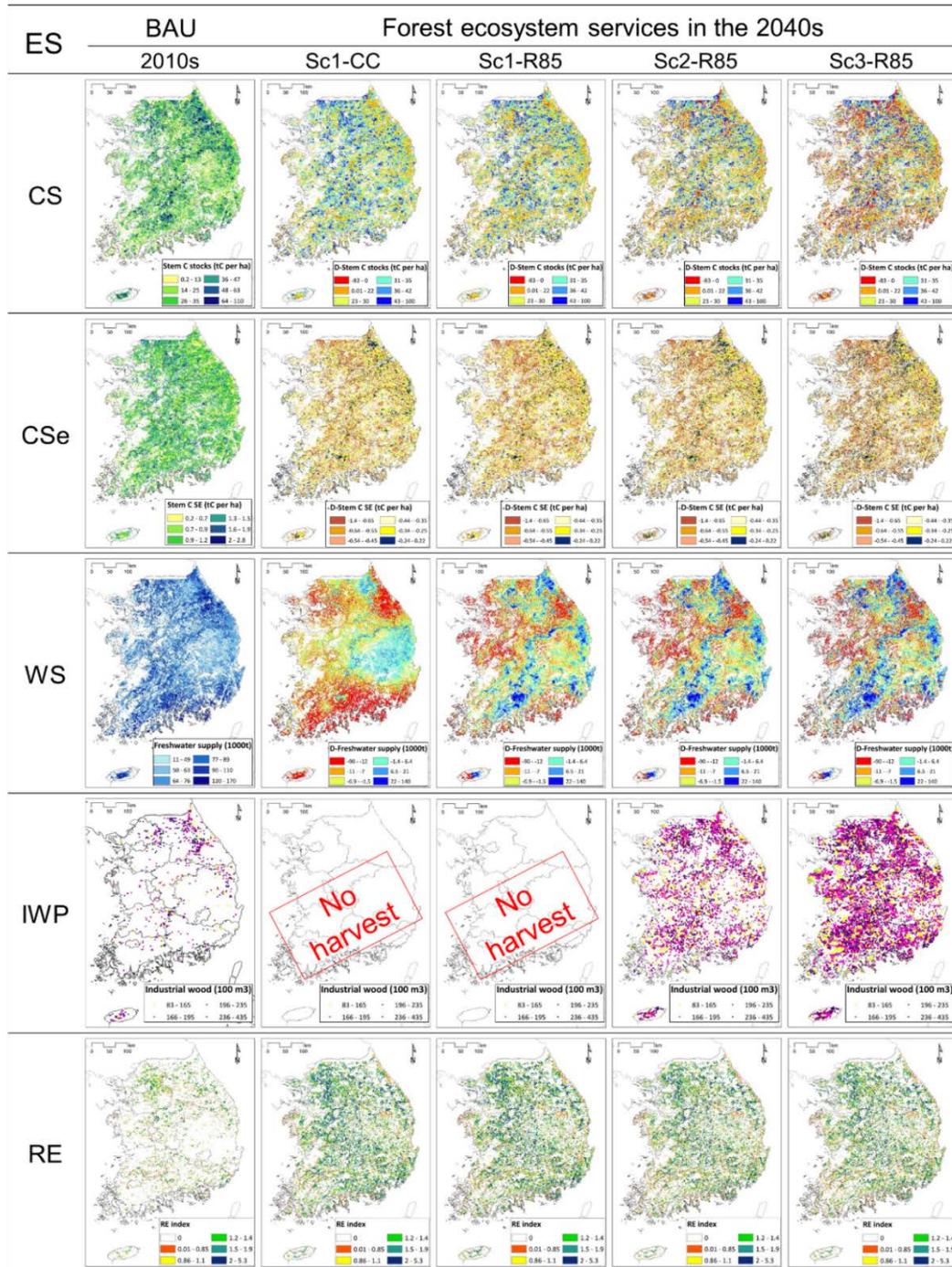
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702

703 **Fig. S5.** Framework for assessing the temporal variation and spatial scales of trade-offs and  
 704 synergies among multiple ecosystem services.

705



706

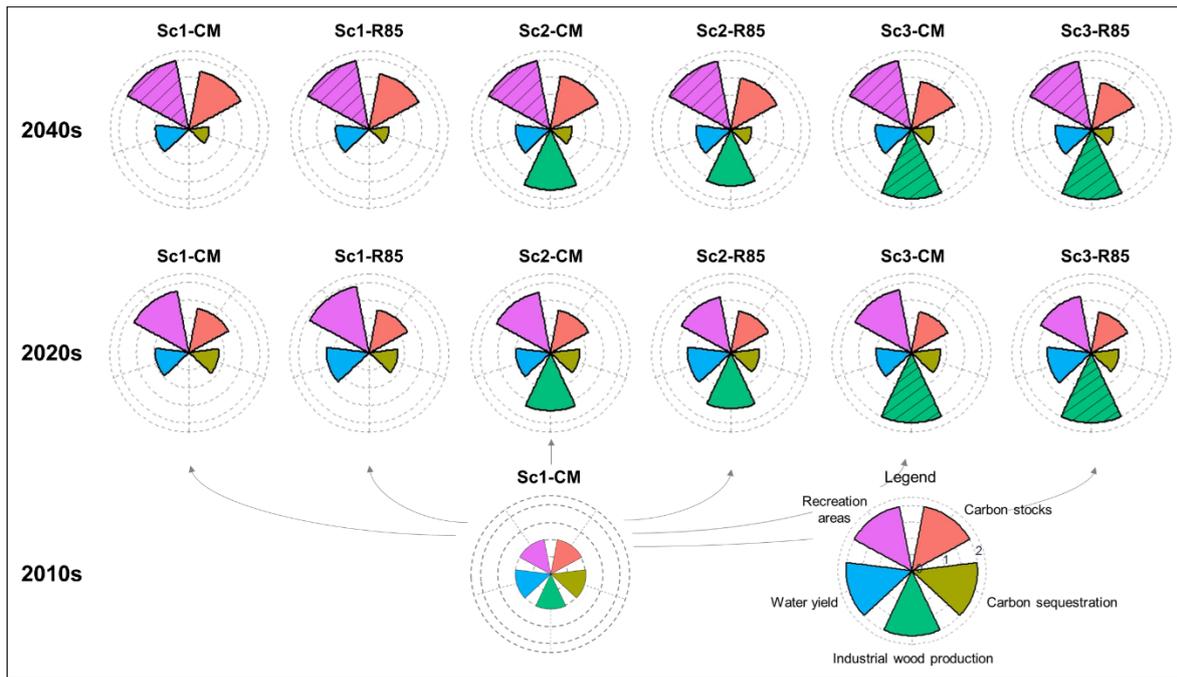
707 **Fig. S6.** The difference in forest ecosystem services (CS, Cse, and WS) between each scenario

708 in the 2040s and in the 2010s. Time and space distribution of IWP and RE in the forest areas

709 of South Korea from the 2010s to the 2040s (CS: forest carbon storage, Cse: annual C

710 sequestration, WS: fresh water supply, IWP: industrial wood production, RE: recreation

711 service).



712

713 **Fig. S7.** Flower diagrams. Estimated changes among forest ecosystem services by each  
 714 scenario in the 2020s and 2040s from the services in the 2010s (as % of 2010s values). Values  
 715 with slashes exceed 200%. Diagrams reveal the percentages for each forest ecosystem service  
 716 through petal length, with each flower representing one scenario in a certain decade.