



22 **Keywords:** Virtual forest land, land use displacement, climate vulnerability, climate

23 change, global supply chains, sustainable development goals

24

## 25 **Introduction**

26 Forest land, occupying more than 31% of global total land area (WB, 2019), is not  
27 only essential for biodiversity conservation (Gibson et al., 2011; Van de Perre et al.,  
28 2018) and climate regulation (Canadell and Raupach, 2008; Murdiyarso et al., 2015),  
29 but also plays a significant role in improving food security (Hickey et al., 2016; Mbow  
30 et al., 2014) and reducing rural poverty (Mohammad Abdullah et al., 2016; Oldekop et  
31 al., 2019). However, global forest area is declining rapidly, especially in the tropics and  
32 subtropics (Hansen et al., 2013). Deforestation from agricultural encroachment has  
33 been widely concerned (Austin et al., 2019; Curtis et al., 2018; Ordway et al., 2019),  
34 yet forest degradation from forestry has so far received relatively less attention. As a  
35 matter of fact, forest degradation can match or even surpass deforestation (e.g. in the  
36 Amazon Basin) (Foley et al., 2007; Hosonuma et al., 2012). Therefore, a holistic profile  
37 of forest land use is pivotal for the conservation and restoration of degraded forest  
38 ecosystems.

39 Local exploitation of forest land is usually driven by external demand in the  
40 context of globalization (Meyfroidt et al., 2013). Since the production and consumption  
41 are increasingly spatially separated and connected through global supply chains (WTO,  
42 2019), an economy can indirectly utilize distant forest land by importing finished forest-  
43 based products. Due to forest scarcity and environmental concerns, an increasing  
44 number of countries tend to displace forest land use outside their borders, putting great  
45 burden on Africa, Southeast Asia and South America (Pendrill et al., 2019), which may

46 be the underlying reason for the forest cover increase among many high- and middle-  
47 income countries along with the rife decrease within low-income countries (FAO, 2016).  
48 As a consequence, regional efforts to preserve forests can be offset by overexploitation  
49 elsewhere, highlighting the necessity to uncover the driving effects of global  
50 consumption and trade on regional forest land exploitation.

51 The area of forest land use displacement is not the only factor that matters, it is  
52 also significant to unveil the climate vulnerability of producing regions. Deforestation  
53 and forest degradation following forest land use have incurred substantial influence on  
54 regional and global climate (e.g. due to changes in carbon and hydrological cycles)  
55 (Malhi et al., 2008; Shukla et al., 1990) while climate change in turn has threatened  
56 forest ecosystems through varying temperature, precipitation, frequency and severity  
57 of natural hazards (Kirilenko and Sedjo, 2007; Lindner et al., 2010). These variabilities  
58 not only aggravate the instability of local production but also undermine the resilience  
59 of global supply chains of forest-based products. Moreover, the influence can extend  
60 beyond forestry to other life-supporting aspects, including food, water, health,  
61 ecosystem services and so on (Sonwa et al., 2012). The vulnerability to climate change  
62 and the readiness to take adaptation actions varied greatly from country to country, due  
63 to diverse ecological, economic, governance and social environments (Chen et al., 2015;  
64 Réjou-Méchain et al., 2021). Therefore, the eco-environmental and socio-economic  
65 consequences of the displacement of the same area of forest land use can be  
66 distinctively different. Considering the pressing and persistent issues of climate change,

67 trade analysis integrating climate vulnerability and adaptation readiness is especially  
68 crucial to help supply chain agents recognize the far-reaching impacts of their  
69 production/consumption and take precautions against potential perturbations through  
70 global mutual efforts, which facilitate both the resilience of global supply chains and  
71 the achievement of multi-dimensional sustainable development goals.

72 The notion of virtual land (aka. land footprint), defined as the total land area  
73 required to produce a kind of good or service, has been widely adopted to measure  
74 different types of land use (Guo et al., 2019; Kan et al., 2021; Kastner et al., 2014),  
75 whereas relatively less efforts have been devoted to analyzing virtual forest land.  
76 Several studies analyzed global virtual timber trade (Kastner et al., 2011; Meyfroidt et  
77 al., 2010; Zhang et al., 2020) and Weinzettel et al. estimated virtual forest land trade  
78 based on the yields of forest products (Weinzettel et al., 2013), but there may be large  
79 uncertainties when converting yields to land use. Yu and colleagues used data on forest  
80 land use directly to assess the tele-connection between regional consumption and global  
81 land use (Yu et al., 2013), while both Weinzettel et al. and Yu et al. concentrated on the  
82 panorama of total land use in a single year. Specific analyses on forest land use and its  
83 temporal trends are still lacking. Moreover, none of previous virtual timber/forest land  
84 studies have considered regions' climate vulnerability and adaptation readiness.

85 Given all of the above, this paper endeavors to trace forest land use across global  
86 supply chains for the world economy, with a specific focus on climate vulnerability and  
87 adaptation readiness of supply chain agents. Temporal trends of forest land use during

88 2000-2015 are investigated and a comprehensive picture for 2015 is presented. By  
89 providing a systematic overview of virtual forest land use, this paper aims to provide  
90 data support and policy implications for policy makers to make informed decisions  
91 regarding both sustainable forest management and improvement of supply chain  
92 resilience without interfering the achievement of other sustainable development goals.

93

## 94 **2. Methods and data sources**

### 95 **2.1 Embodiment accounting**

96 Virtual forest land is calculated by means of embodiment accounting, based on  
97 input-output analysis (IOA), which has been applied to study materials (Bruckner et al.,  
98 2012; Giljum et al., 2015; Wiedmann et al., 2015), total primary energy (Chen and Wu,  
99 2017; Kan et al., 2019a; Moreau and Vuille, 2018), oil (Wu and Chen, 2019), natural  
100 gas (Kan et al., 2020; Kan et al., 2019b), coal (Wu and Chen, 2018), water (Ewing et  
101 al., 2012; Feng et al., 2014; Guan and Hubacek, 2007), carbon emissions (Feng et al.,  
102 2015; Feng et al., 2013; Guan et al., 2018) and so on. Leontief is regarded as the pioneer  
103 to integrate environmental accounts into IOA (Leontief, 1970). Bullard and Herendeen  
104 adapted IOA to calculate energy embodied in goods and services for the United States,  
105 on the basis of an energy balance model (Bullard and Herendeen, 1975a, b; Bullard et  
106 al., 1978). By integrating the embodiment theory in systems ecology (Odum, 1983), the  
107 model was further extended by Chen and colleagues to study various ecological  
108 elements (Kan et al., 2020; Liu et al., 2020; Wu et al., 2018; Wu et al., 2019).

109 From the biophysical perspective, practically everything in the earth can be  
110 considered as a direct or indirect product of the energy and materials that underpin its  
111 formation. Industrial capital, intermediate and final products created by the economic  
112 process are all maintained by exogenous environmental input (Costanza, 1980). The  
113 embodiment theory was therefore developed to probe into the biophysical support of  
114 the economic system (Odum, 1983, 1995; Wu et al., 2018). Just as every commodity is  
115 regarded to have economic value measured by currency in the monetary accounting  
116 framework, the theory reveals that every product embodies the resource use required to  
117 produce it, i.e. virtual resource use, no matter whether the product is for intermediate  
118 or final use. Another basis of the embodiment accounting is the objective law of  
119 conservation of resource use, such as conservation of embodied energy (Bullard and  
120 Herendeen, 1975a, b; Hannon et al., 1983; Herendeen, 2004). In other words, the total  
121 input of virtual resource use to a system is equal to the system's total output. The  
122 biophysical balance can be applied to every system, including each economic sector.

123 In this paper, the embodiment accounting is carried out with the help of the multi-  
124 regional input-output model, which simulates the world as a  $m$  region  $\times$   $n$  sector  
125 economic system and captures transactions between sectors in intermediate trade and  
126 sectoral sales to final consumers in final market. Once forest land is exploited, it enters  
127 the economic system and exists as product embodiments. Though the physical forest  
128 land is immovable, the utility of forest land can be transferred along with currency flows  
129 and goods/services flows (Hilten, 2019), from primary exploiters to final consumers.

130 Sectoral input of virtual forest land consists of forest land exploited directly by the  
 131 sector and that embodied in imported intermediate products, while the total output is  
 132 comprised of forest land use embodied in all the sector's production. Then the  
 133 biophysical balance equation for sector  $i$  in region  $r$  can be written as:

$$fl_i^r + \sum_{s=1}^m \sum_{j=1}^n (\varepsilon_j^s z_{ji}^{sr}) = \varepsilon_i^r \left( \sum_{s=1}^m \sum_{j=1}^n z_{ij}^{rs} + \sum_{s=1}^m f_i^{rs} \right) \quad (1)$$

134 where  $fl_i^r$  denotes forest land exploited directly by sector  $i$ .  $z_{ji}^{sr}$  represents the  
 135 monetary value of intermediate products supplied by sector  $j$  in region  $s$  to sector  $i$ .  $f_i^{rs}$   
 136 stands for monetary value of final products traded from sector  $i$  to region  $s$ .  $\varepsilon_i^r$  is the  
 137 virtual forest land intensity, representing forest land use embodied in the sector's unit  
 138 output. Considering the biophysical equations for all the sectors, equation (1) can be  
 139 transformed as:

$$\mathbf{FL} + \boldsymbol{\varepsilon} \mathbf{Z} = \boldsymbol{\varepsilon} \mathbf{X} \quad (2)$$

140 where  $\mathbf{FL} = (fl_i^r)_{1 \times mn}$ ,  $\mathbf{Z} = (z_{ij}^{rs})_{mn \times mn}$ ,  $\boldsymbol{\varepsilon} = (\varepsilon_i^r)_{1 \times mn}$  and  $\mathbf{X} = \text{diag}(x_i^r)_{mn \times mn}$   
 141 ( $x_i^r = \sum_{s=1}^m \sum_{j=1}^n z_{ij}^{rs} + \sum_{s=1}^m f_i^{rs}$ ). Therefore,  $\boldsymbol{\varepsilon}$  can be calculated as:

$$\boldsymbol{\varepsilon} = \mathbf{FL}(\mathbf{X} - \mathbf{Z})^{-1} \quad (3)$$

142 The amount of forest land use embodied in the products provided by sector  $i$  in  
 143 region  $r$  can be calculated as the monetary value of the products multiplied by  
 144 corresponding intensity  $\varepsilon_i^r$ . Then virtual forest land use associated with final  
 145 consumption and international trade can be obtained by summing up forest land use  
 146 embodied in all the consumed or traded products. Therefore, for region  $r$ , virtual forest



147 land use in its final consumption ( $VF^r$ ), its intermediate imports ( $VFI_{int}^r$ ) and exports  
 148 ( $VFE_{int}^r$ ), final imports ( $VFI_{fin}^r$ ) and exports ( $VFE_{fin}^r$ ) can be expressed as:

$$VF^r = \sum_{s=1}^m \sum_{j=1}^n \varepsilon_j^s f_j^{sr} \quad (4)$$

$$VFI_{int}^r = \sum_{s=1(s \neq r)}^m \sum_{j=1}^n \sum_{i=1}^n (\varepsilon_j^s z_{ji}^{sr}) \quad (5)$$

$$VFE_{int}^r = \sum_{s=1(s \neq r)}^m \sum_{j=1}^n \sum_{i=1}^n (\varepsilon_i^r z_{ij}^{rs}) \quad (6)$$

$$VFI_{fin}^r = \sum_{s=1(s \neq r)}^m \sum_{j=1}^n (\varepsilon_j^s f_j^{sr}) \quad (7)$$

$$VFE_{fin}^r = \sum_{s=1(s \neq r)}^m \sum_{i=1}^n (\varepsilon_i^r f_i^{rs}) \quad (8)$$

149

## 150 2.2 Data sources

151 The input-output tables are taken from EXIOBASE 3, due to its high product  
 152 resolution (i.e. 200 products with products of forestry, logging and related services  
 153 separated from agricultural products) matched with land use data for multiple land  
 154 covers (including cropland, grazing land and forest land) (Stadler et al., 2018). It should  
 155 be noted that, the end year of land extensions in original EXIOBASE 3 data series is  
 156 2011, so statistics of land use are collected from EXIOBASE 3rx, which provides land  
 157 updates to 2015 for 214 countries and regions. EXIOBASE 3 assumes marginally used  
 158 areas (grid cells on the map that contain both used and unused land) as uses with low  
 159 intensities (e.g. hunting) and does not attribute this part to forestry operations. This

160 paper only considers forest land use by forestry sectors. With reference to *Global Forest*  
161 *Products – Facts and Figures* (FAO, 2018a), *Yearbook of Forest Products* (FAO, 2019b)  
162 and *Forestry Production and Trade* statistics (FAO, 2019a), forest products in this paper  
163 refer to roundwood and various derived products, such as sawn wood, veneer sheets,  
164 wood-based panels, wood pulp, paper, paperboard, wood fuel, wood chips, wood  
165 particles and wood pellets. Non-forest products refer to other products that are  
166 indirectly associated with forest land exploitation.

167 The world is aggregated into 44 major economies and 5 *rest of world* (RoW)  
168 composite regions in EXIOBASE 3 (please see the Supplementary Information for the  
169 classification). Some of the regions are further aggregated in this paper for research  
170 purpose. The EU includes 28 member countries, China includes EXIOBASE 3  
171 categories of *China* and *Taiwan, China*. To distinguish from EXIOBASE 3 *RoW Middle*  
172 *East* and *RoW Africa* categories, Middle East covers *Turkey* and *RoW Middle East* and  
173 Africa covers *South Africa* and *RoW Africa*. Population data are derived from the World  
174 Bank (WB, 2017).

175 Data on climate vulnerability and adaptation readiness are taken from University  
176 of Notre Dame Global Adaptation Index (ND-GAIN) Country Index (Chen et al., 2015).  
177 Vulnerability reflects the propensity to be adversely affected by climate disruption,  
178 which is measured by sensitivity, exposure and adaptive capacity. Readiness assesses  
179 the ability to leverage public and private investments to take adaptation actions in the  
180 face of climate hazards, which is measured by economic, governance and social

181 readiness. Due to its inclusiveness, feasibility and comparability, ND-GAIN Country  
182 Index and similar conceptual framework have been applied in several studies (Lindner  
183 et al., 2010; Scarano and Ceotto, 2015; Scheelbeek et al., 2020; Sonwa et al., 2012).  
184 Countries are assigned to 8 categories based on their vulnerability scores or readiness  
185 scores. With reference to Scheelbeek et al. (Scheelbeek et al., 2020), countries with the  
186 highest to the fourth highest levels are considered to have relatively high climate  
187 vulnerability / low adaptation readiness. This paper uses average scores over 2000-2015  
188 for classification.

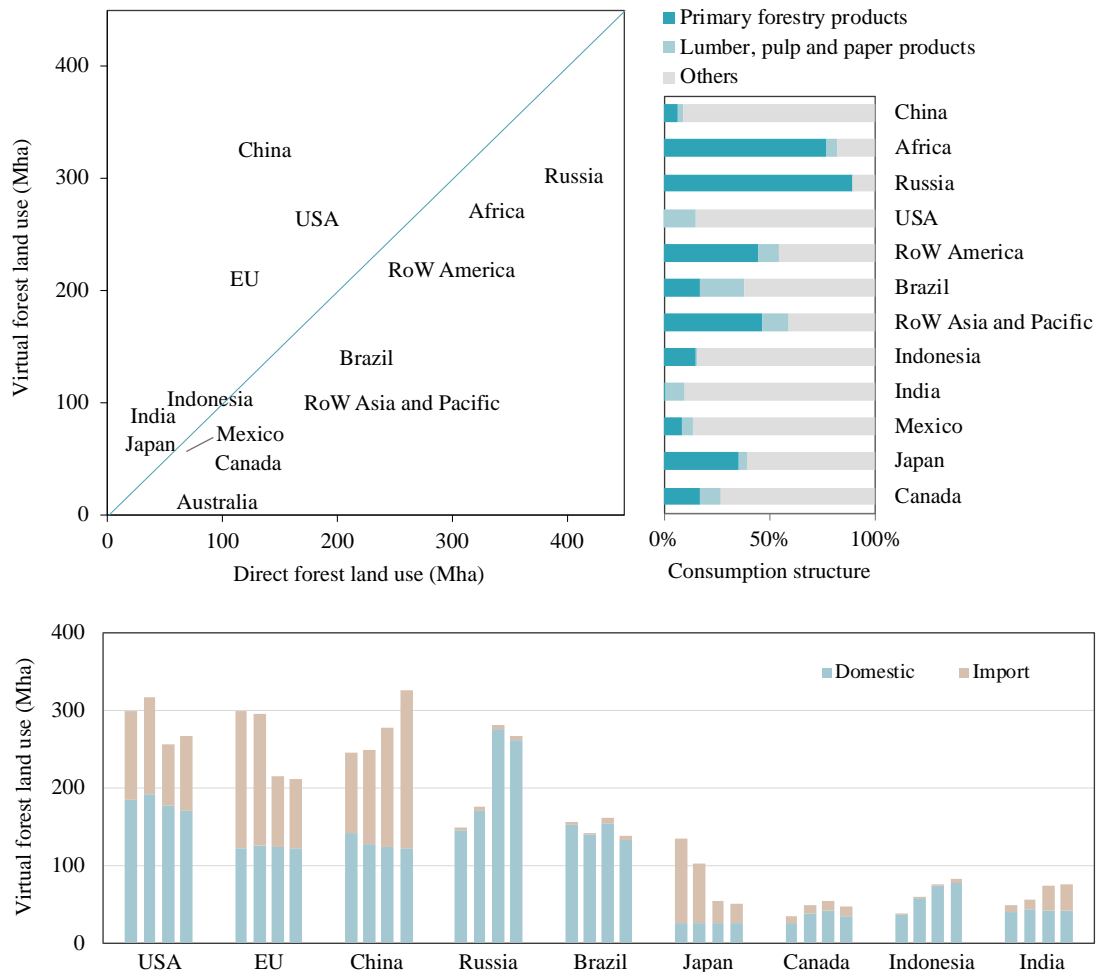
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### 190 **3. Results**

#### 191 **3.1 Virtual forest land use associated with final consumption**

192 At a global scale, forest land use amounted to 2268.3 million hectares (Mha) in  
193 2015. As shown in Fig. 1, China was the leading final consumer of virtual forest land  
194 use (325.7 Mha), followed by Africa (292.7 Mha), Russia (266.4 Mha), the USA (266.1  
195 Mha), RoW America (218.8 Mha) and Brazil (138.1 Mha). Virtual forest land use of  
196 other countries and regions each were less than 100 Mha. For China, the EU, the USA,  
197 India and Japan, forest land use associated with final consumption was greater than  
198 local forest land exploitation, which means a large amount of forest land use was  
199 displaced to other regions. The situation was reversed for Russia, Brazil, Canada,  
200 Australia, Mexico and many other African, South American and Asia-Pacific countries,  
201 which manifests their role as suppliers who exploited local forests for downstream use

202 both at home and abroad.



203  
204 **Fig. 1 Virtual forest land use in final consumption**

205 The upper left chart compares direct and virtual forest land use in 2015 for EXIOBASE 3 countries and  
206 regions as well as EU members as a whole, with the size of the circles denoting per capita virtual forest  
207 land use. The upper right chart presents sectoral contribution to virtual forest land use for dominant  
208 final consumers. The bottom chart shows the temporal trends of virtual forest land use by source for  
209 major final consumers (from left to right are results for 2000, 2005, 2010 and 2015).

210 Only in Russia, Africa, RoW Asia and Pacific and RoW America, traditional  
211 primary (e.g. raw round wood and wood chips ) and processed forest products (e.g.  
212 lumber, pulp and paper) represented over 50% of virtual forest land use (89%, 82%, 59%  
213 and 54%, respectively), while they comprised a small portion in other dominant  
214 consumer countries: less than 10% in China and India and less than 16% in the USA,

215 Indonesia and Mexico. This is because a considerable share of forest products were  
216 used as industrial intermediate input to produce other highly-processed products and to  
217 support tertiary sectors.

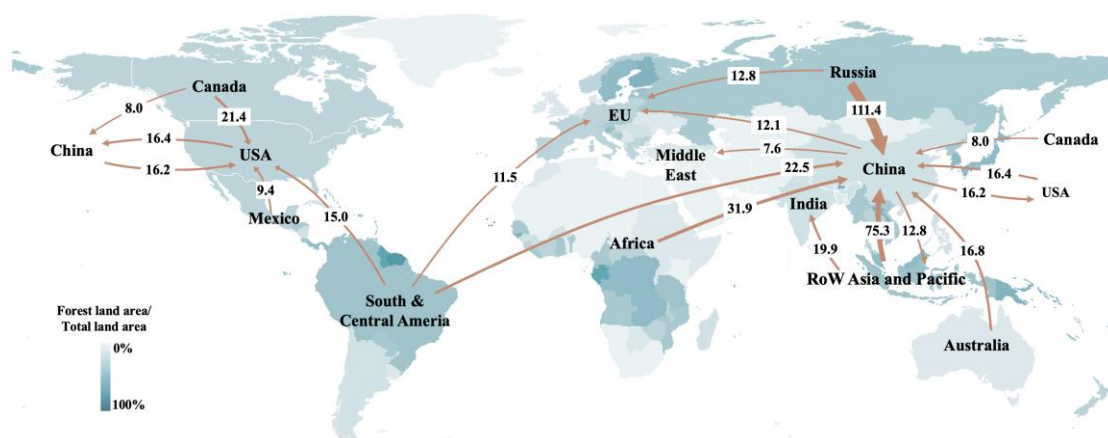
218 Fig. 1 also presents temporal trends of virtual forest land use for dominant final  
219 consumers. Generally, developed economies (including the USA, the EU and Japan but  
220 excluding Canada) witnessed decreasing consumption, but developing economies  
221 (including China, Russia, Indonesia and India) exhibited growing demand. In 2000, the  
222 USA and the EU were the top 2 final consumers (around 300 Mha each), but China and  
223 Russia took their positions in 2015. Virtual forest land use of Brazil fluctuated slightly  
224 around 150 Mha and that of Canada showed a small inverted-U shape. Moreover, the  
225 USA, the EU, China, Japan and India remained major importers during this period, and  
226 changes in virtual forest land use were mainly due to changes in import volume.

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### 228 **3.2 Virtual forest land use associated with international trade**

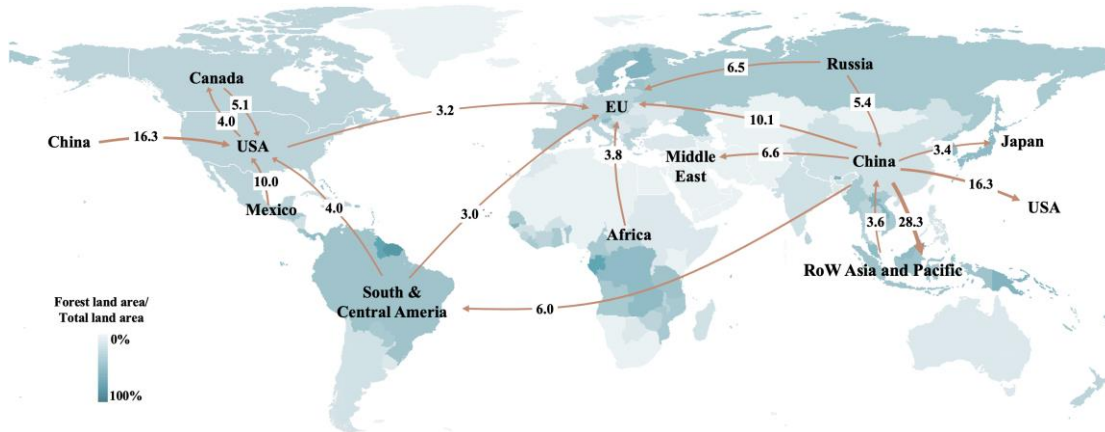
229 Fig. 2 shows major trade flows in intermediate and final trade in 2015, in order to  
230 reveal how countries were interconnected in the intricate global trade network to  
231 eventually deliver finished products to final consumers. Forest land use embodied in  
232 trade among EXIOBASE 3 regions amounted to 41% of global total forest land  
233 exploitation in magnitude, with 74% arising from intermediate trade. In intermediate  
234 trade, Russia, Africa, Latin America and RoW Asia and Pacific were major exporters  
235 of forest-risk products, while China, the USA and the EU were dominant markets.

236 China was the largest importer with diversified trade partners: Russia, RoW Asia and  
 237 Pacific, Africa and South & Central America respectively contributed 111.4 Mha, 75.3  
 238 Mha, 31.9 Mha and 22.5 Mha . Virtual forest land use embodied in the USA’s import  
 239 was much less than that of China, dominantly associated with export from Canada (21.4  
 240 Mha), China (16.2 Mha), South & Central America (15.0 Mha) and Mexico (9.4 Mha).  
 241 Virtual forest land use linked to the EU’s import derived primarily from Russia (12.8  
 242 Mha), China (12.1 Mha), and South & Central America (11.5 Mha). India was also an  
 243 significant importer, with large amounts of virtual forest land use originated from RoW  
 244 Asia and Pacific (19.9 Mha). Most leading importers and exporters in intermediate trade  
 245 held their positions in final trade, while China turned to be the largest exporter,  
 246 associated with 28.3 Mha of virtual forest land exported to RoW Asia and Pacific, 16.3  
 247 Mha to the USA and 10.1 Mha to the EU, which confirms China’s role as the world  
 248 factory.



249  
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(a)



(b)

Fig. 2 Forest land use embodied in dominant trade flows in (a) intermediate trade and (b) final trade

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### 256 3.3 Source to sink budget

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Virtual forest land traded from primary suppliers to final consumers usually flows

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through at least one intermediate producer. It is still unknown how much burden is put

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on a certain primary supplier (source) to satisfy the final consumption of a certain final

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consumer (sink). The source-to-sink budget is therefore required to identify the

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connection between source and sink regions. For EXIOBASE 3 regions, 672.6 Mha of

262

forest land was linked to final consumption elsewhere in 2015, reaching 30% of global

263

total exploitation. As shown in Fig. 3, 35% (142.8 Mha) of forest land exploitation in

264

Russia was associated with export production, with China and the EU accounting for

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77.1 Mha and 22.6 Mha, respectively. For Indonesia, Australia and RoW Asia and

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Pacific (as an aggregate region), 42% (128.1 Mha) was displaced from other regions,

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mainly from China (61.9 Mha) and India (19.1 Mha). The proportions for Brazil and

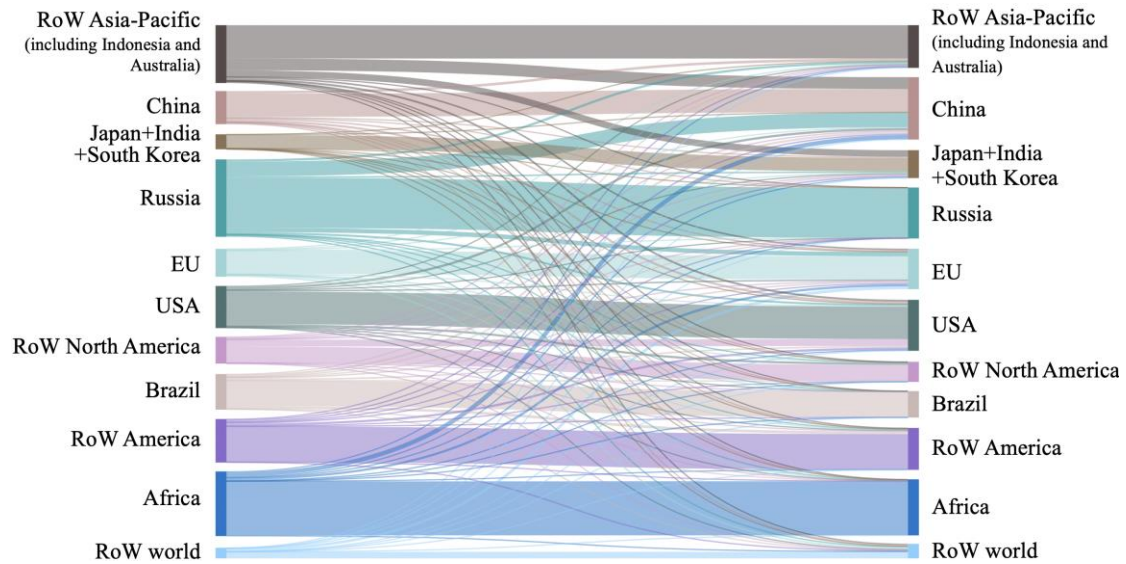
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RoW America were 28% (52.7 Mha) and 19% (42.1 Mha) respectively, with China, the

269

USA and the EU each making up 14 Mha - 20 Mha. The ratios for Canada and Mexico

270 reached 52% (37.4 Mha) and 28% (19.0 Mha), and the USA alone was responsible for  
 271 a quarter (35.7 Mha). Forest land use displaced to Africa also reached 55.4 Mha, but  
 272 only comprised 16% of local forest land use.



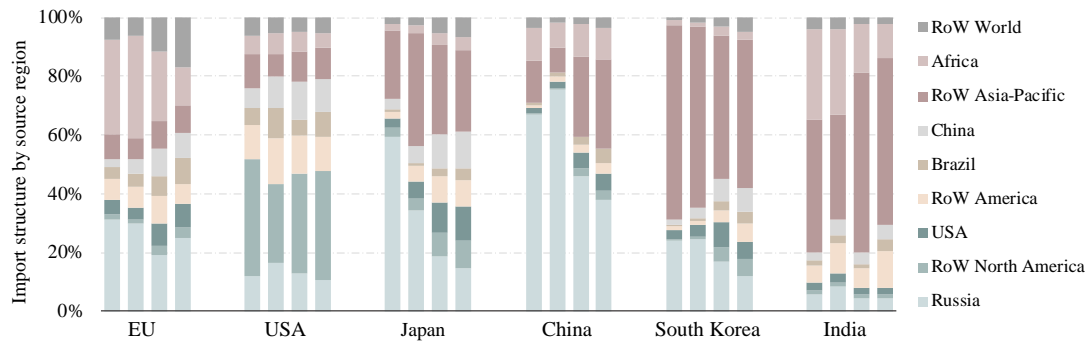
**Fig. 3 Source-to-sink budget of global forest land use**

273 The world is divided into 11 regions. The left side of the Sankey diagram shows direct forest land use  
 274 in source regions, and the right side shows forest land use embodied in the final consumption of sink  
 275 regions. The thickness of the flows (using source regions' colors) denotes the amount of forest land  
 276 exploited by source regions to satisfy the final consumption of sink regions.  
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 278

279 Forest land exploited in the regions categorized to the highest - 4<sup>th</sup> highest climate-  
 280 vulnerable levels rose from 634.1 Mha in 2000 to 689.4 Mha in 2015, accounting for  
 281 over half of total increase in global forest land use. Forest land use displaced from China  
 282 to these climate-vulnerable regions rose from 24.1 Mha in 2000 to 63.8 Mha in 2015  
 283 and that displaced from India rose from 6.0 Mha to 19.5 Mha, due to escalating imports  
 284 from tropical and subtropical regions (mainly Asia-Pacific). For the EU and the USA,  
 285 despite the decreasing import volumes, displacement to climate-vulnerable regions  
 286 remained relatively large in magnitude (21.2 Mha and 16.8 Mha, respectively).  
 287 Meanwhile, for most source regions, the readiness to take adaptation actions was also

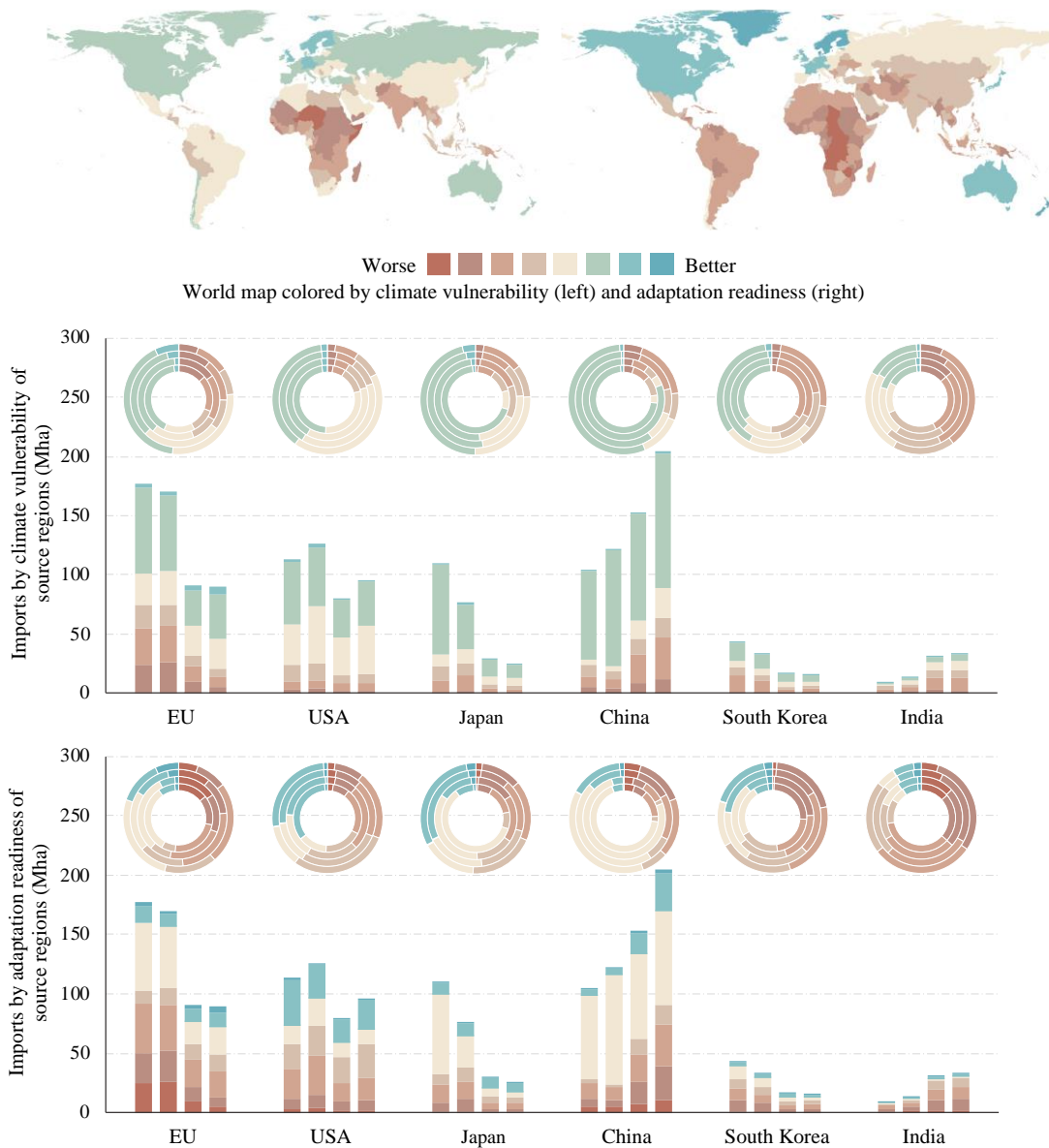


288 low. Consequently, the proportion displaced to the least – 4<sup>th</sup> least ready regions  
 289 reached 50% - 90% for the EU, the USA, South Korea and India while climbed up from  
 290 27% to 44% for China and from 30% to 51% for Japan.



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 293

**Fig. 4 Import structure by source region for major importers in 2000, 2005, 2010 and 2015 (from left to right)**



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295 **Fig. 5 Imports by climate vulnerability and adaptation readiness of source regions**  
 296 **for major importers in 2000, 2005, 2010 and 2015 (from left to right)**

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298

299 **4. Discussion and policy implications**

300 The extensive displacement of forest land use reflects the globalization of forest

301 land use facilitated by economic globalization. On the one hand, a large share of forest

302 land use was displaced from forest-scarce places to forest-rich places or between places

303 with different forest types. All the agents participating in the global supply chains can

304 take what they want: importers can avoid forest exploitation and improve economic  
305 efficiency by taking advantage of high-yield and high-quality forests elsewhere, and  
306 meanwhile the exporters can gain national revenue by selling forest products. On the  
307 other hand, the land use was generally displaced from developed (e.g. the EU, the USA  
308 and Japan) and emerging (e.g. China and India) economies to developing ones (e.g.  
309 Russia, Africa, South America, South and Southeast Asia), partly because stricter  
310 standards and higher costs in high-regulating economies cause forest exploitation  
311 spillover to lower-regulating jurisdictions (Rulli et al., 2019). As a consequence, regions  
312 with high climate vulnerability and low adaptation readiness are facing growing  
313 pressure, which may exacerbate forest loss at a global scale. Actually, forestry is found  
314 to be the leading proximate cause of global intact forest loss (Potapov et al., 2017a).  
315 Even selective logging, which is often regarded as sustainable, is fragmenting the  
316 world's most precious tropical forests in the Amazon and Congo Basin (Cazzolla Gatti  
317 et al., 2015). These impacts can extend from forestry to other life-supporting socio-  
318 economic aspects. For instance, over 800 million people living below poverty line in  
319 the tropics reside around forests and savannahs, who rely on the ecosystems to meet  
320 their energy, water, nutritional, medicinal and cultural needs (FAO, 2018b; Sonwa et  
321 al., 2012). The influence can be also transmitted from producers to downstream supply  
322 chain agents. As forestry production can be highly variable in case of droughts, floods,  
323 wildfires, wind storms and insect outbreaks due to climate change (Kirilenko and Sedjo,  
324 2007), heavy reliance on climate-vulnerable countries would pose a threat to the

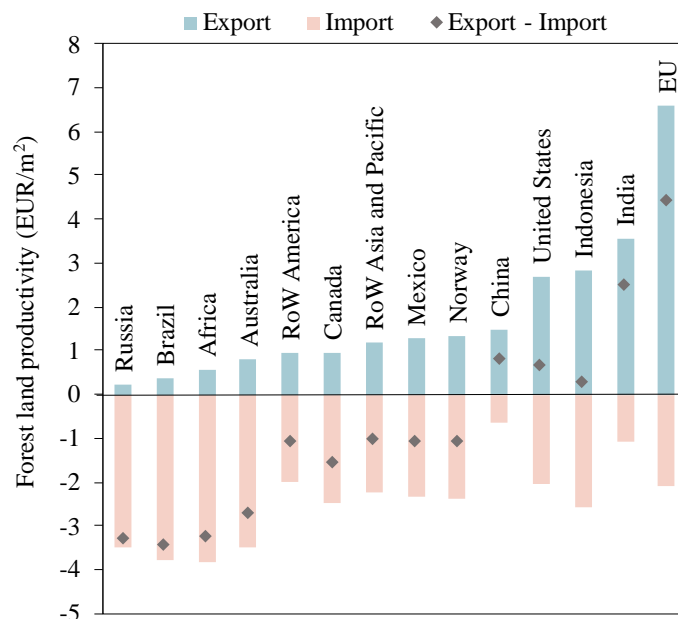
325 resilience of supply chains of forest-based products. The EU, the USA, China and India  
326 should pay particular attention, given their substantial virtual forest land import from  
327 Latin America, Africa, Southeast Asia and Oceania.

328         Current forest policies and programs, such as US Healthy Forests Restoration Act,  
329 Natural Forest Conservation Program of China, EU Natura 2000 and Brazilian Forest  
330 Act, mostly target on local forest conservation. However, stringent regulation in  
331 production regions restricts local exploitation without suppressing downstream demand,  
332 and the consequent supply shortage may spark illegal exploitation and trade, while  
333 forest protection in consumption regions alone may aggravate land use displacement.  
334 Therefore, transnational joint efforts are needed to ameliorate negative impacts caused  
335 by land use globalization. From one aspect, deforestation and forest degradation from  
336 land use should be mitigated. Given economies are connected through international  
337 trade, market-oriented approaches can be adopted. Production regions can impose taxes  
338 on exploiters for ecosystem services loss and carbon emissions, so that environmental  
339 and ecological externalities can be internalized. The tax can be also earmarked for  
340 sustainable forest management specifically. Dominant final consumers, such as the  
341 USA, China, the EU, Japan and India, can adopt forest certification, which promotes  
342 management of distant forests by informing consumers whether the production process  
343 conforms to certain standards and thus helping weed out products that are not eco-  
344 friendly (Villalobos et al., 2018). From another aspect, it is imperative to help climate-  
345 vulnerable producers to adapt to climate perturbation. Developed economies can share

346 successful forest management experiences and technologies, such as fire controlling via  
347 regulation and education and satellite-based monitoring system that can guarantee  
348 traceability (Gralewicz et al., 2012; Verhegghen et al., 2016). It is also helpful to  
349 provide experiences of enhancing institutional and society arrangement that can  
350 effectively leverage investments for adaptation actions, since insufficient financial  
351 resource is a major barrier to forest management in many regions (Malhi et al., 2008).  
352 For example, recent years have witnessed surging foreign investments in Africa  
353 (Conigliani et al., 2018), then the market forces can be directed to improve local  
354 adaptation capacity in case of climate disruption. Moreover, in order to strengthen  
355 supply chain resilience, consumers can diversify or reduce reliance on imports,  
356 especially those from climate-vulnerable regions.

357 Above discussions of forest land use are from the perspective of forests' provision  
358 of ecosystem services and material resources, while forest land also serves other socio-  
359 economic functions, such as creating national revenue. According to the World Bank,  
360 forest rents contributed 2.4% of GDP in Sub-Saharan Africa in 2011 and the ratio was  
361 even higher in individual African countries (WB, 2018). To preliminarily measure the  
362 economic efficiency of forest land use, an indicator termed as forest land productivity  
363 is constructed, with reference to the indicator "resource productivity" adopted by  
364 OECD (OECD, 2011). It is calculated as monetary trade volume divided by forest land  
365 use embodied in corresponding trade flows. As shown in Fig. 6, forest land  
366 productivities associated with export of most production regions were low, reflecting

367 their specialization in low value-added segments of global supply chains. Poor  
 368 economic return may stimulate more forest exploitation, which in turn impairs land  
 369 productivity and leads to a vicious circle. Hence, these regions should expand capacity  
 370 to add value to forest-risk products, gradually getting rid of the so-called resource curse.  
 371 Moreover, there may be conflicts between eco-environmental and socio-economic  
 372 functions of forest land. For example, forestry can trigger negative environmental  
 373 impacts, while strict logging prohibition may cut off the already pitiful income of local  
 374 residents. The uneven distribution of gains and losses will further lead to social  
 375 inequality and instability. Therefore, future work should pay more attention to the  
 376 synergies and trade-offs between multiple functions of forest land.



377  
 378 **Fig. 6 Forest land productivity associated with import and export for dominant**  
 379 **supply chain agents**

380 Our results also have academic implications regarding the accounting of virtual  
 381 forest land use. A substantial share of forest land use was associated with the final

382 consumption of non-forest products. In other words, countries can make use of foreign  
383 forest land even without importing traditional forest products. Conventional trade  
384 balance analyses of only forest products will overestimate forest land use driven by  
385 manufacturing countries (usually developing countries) while underestimating that of  
386 actual final consumers. Given the growing awareness to adopt consumption side  
387 policies and supply chain initiatives, it is necessary to identify the real end users and  
388 reveal their roles in driving forest land use. Due to the intricate international trade  
389 network, it is increasingly difficult to trace forest land use through the entire global  
390 supply chains with tools based on physical trade flows. The embodiment accounting  
391 that inclusively covers all the trade flows can provide complementary information to  
392 existing analyses that only consider traditional forest products.

393

#### 394 **Limitations and future perspectives**

395 Data on direct forest exploitation have uncertainties. There is no agreed-upon  
396 definitions of forests. For example, FAO defines forest land as an area more than 0.5  
397 hectare with trees higher than 5 meters and tree cover greater than 10% (FAO, 2010;  
398 Schepaschenko et al., 2015), while others set a threshold of 20% (Potapov et al., 2017b)  
399 or 25% (Pendrill et al., 2019). This may lead to the discrepancies in the findings of  
400 different studies. There are also limitations in input-output analysis resulting from the  
401 compilation of input-output tables. Due to the regional aggregation in EXIOBASE 3  
402 input-output tables, forest land use displaced via international trade is underestimated.

403 It is also hard to identify the actual importers and exporters at a subnational scale.  
404 Physical and hybrid accounting are alternative tools. However, physical accounting has  
405 limits in identifying the origin and end users and there are large uncertainties to convert  
406 yield of forest products to actual land use (also the problem of hybrid approach that  
407 combines IOA and physical accounting) (Bruckner et al., 2015; Weinzettel et al., 2013).  
408 Since this paper focuses on the displacement of forest land use, the capability of input-  
409 output analysis to trace forest land use from the origin to the end is especially important  
410 to us. Considering current data availability and technical feasibility, analyses with  
411 different methods and diverse focuses are essential to provide complementary  
412 information.

413         Meanwhile, several problems remain to be solved in the future. First, it would be  
414 better to consider the heterogeneities of forest land in terms of productivity, vegetation  
415 type, carbon sequestration, biodiversity conservation and other ecosystem services.  
416 Since the concept of forest used in the present and most previous papers includes all  
417 the areas that meet certain thresholds which wipes out the heterogeneities of forests, it  
418 is hard to tell exactly whether the transfer of forest land use bring about net gains or net  
419 losses from different perspectives. Consequently (Second), future analyses should not  
420 be limited to land itself but investigate multiple consequences of forest land use, such  
421 as changes in biodiversity and carbon stock due to land use displacement. Third, more  
422 efforts are needed to improve the region resolution of forest land analysis. This paper  
423 conducts the virtual land accounting at the expense of limited region details, which



424 hinders more detailed analyses for individual countries. Scholars in different fields are  
425 encouraged to strengthen communication and cooperation in order to develop methods  
426 and databases that are more suitable for research with different purposes.

427

## 428 **Conclusions**

429 This paper provides a comprehensive overview of forest land use across global  
430 supply chains for the world economy by means of embodiment accounting based on  
431 the data provided by EXIOBASE 3. A specific focus is put on the climate vulnerability  
432 and adaptation readiness of supply chain agents, based on Notre Dame Global  
433 Adaptation Index (ND-GAIN) Country Index.

434 2268.3 Mha of forest land was exploited for forestry in 2015, with China as the  
435 leading final consumer of virtual forest land use (325.7 Mha), followed by Africa (292.7  
436 Mha), Russia (266.4 Mha), the USA (266.1 Mha), RoW America (218.8 Mha) and  
437 Brazil (138.1 Mha). For EXIOBASE 3 regions, 30% (672.6 Mha) of global total forest  
438 land use was linked to export production. For Russia, the largest exporter, 35% (142.8  
439 Mha) of forest exploitation was embodied in export, with China and the EU accounting  
440 for 77.1 Mha and 22.6 Mha, respectively. The proportions for other major producers  
441 ranged between 16% (Africa) - 52% (Canada), respectively. Developed economies  
442 (including the USA, the EU and Japan but excluding Canada) generally decreased  
443 virtual forest land import, but China and India showed growing reliance on the tropical  
444 and subtropical regions (especially in the Asia-Pacific), which are vulnerable to climate

445 change and have low readiness to take adaptation actions. At a global scale, increasing  
446 forest land was exploited in climate-vulnerable regions, contributing over 50% of the  
447 total increase in global forest land use.

448 Therefore, transnational joint efforts are needed to ameliorate negative  
449 consequences caused by forest land use displacement. On the one hand, forest loss  
450 following forest land use should be mitigated, for example, through ecological  
451 payments, carbon tax and forest certification. On the other hand, consumer countries  
452 should help climate-vulnerable source countries to adapt to climate disruption, for  
453 example, by sharing experiences and techniques to monitor and control climate  
454 disruption and to enhance institutional, society and economic environment for  
455 adaptation actions. Furthermore, synergies and trade-offs between the multiple  
456 functions of forest land in eco-environmental and socio-economic systems should be  
457 considered.

458

## 459 **Declaration of interests**

460 The authors declare no competing interests.

461

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