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eprints@whiterose.ac.uk https://eprints.whiterose.ac.uk/ 1 Title: The extent of forest in dryland biomes

2 One sentence summary:

3 Previously unreported forest areas in dryland biomes increase current estimates of the

4 global forest cover by at least 9 %.

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45 Abstract

46 Dryland biomes cover two fifths of the Earth's land surface but their forest area is 47 poorly known. Here, we report an estimate of global forest extent in dryland biomes, 48 based on analysing more than 210,000 0.5 ha sample plots through a photo-49 interpretation approach using large databases of satellite imagery at (i) very high spatial 50 resolution and (ii) very high temporal resolution which are available through the Google 51 Earth platform. We show that, in 2015, 1,327 million ha of drylands had more than 10% 52 tree-cover, and 1,079 million ha comprised forest. Our estimate is 40-47 % higher than 53 previous estimates, corresponding to 467 million ha of forest that have never been 54 reported before. This increases current estimates of global forest cover by at least 9 %.

55 Main text

56 Dryland biomes cover about 41.5 % of the Earth's land surface (1). They contain some 57 of the most threatened, yet disregarded, ecosystems (2, 3), including seven of the twenty 58 five biodiversity hotspots (4), while facing pressure from climate change and human 59 activity (5, 6). The most recent climate model simulations, based on contrasted 60 Representative Concentration Pathways (RCPs), i.e. RCP 8.5 and RCP 4.5, show that 61 global climate change could cause dryland biomes to expand by 11% to 23% by the end of the 21st century (7). If this occurs, dryland biomes could cover more than half of the 62 63 global land surface (7). Climate change will lead to extended droughts, regional 64 warming (8, 9) and, combined with a growing human population, to an increased risk 65 of land degradation and desertification in the drylands (7). Such changes will 66 particularly affect developing countries, where most dryland expansion is expected to 67 occur (7, 10) and where woody resources provide key goods and services to support 68 human livelihoods (11).

69

70 However, our current knowledge of the extent of tree cover and forests in drylands is 71 limited. This is illustrated by significant spatial disagreements between recent satellite-72 based global forest maps (12-14) and by the scarcity of large-scale studies of dryland 73 biomes (3). The most recent estimates of tropical dry forest extent based on remote 74 sensing surveys vary greatly, from 105 Mha for the year 2000, derived from a wall-to-75 wall map at coarse resolution (5) to 542 Mha for the year 2010 derived from a global 76 sample of medium resolution images (15). This disparity can partly be explained by 77 differences in satellite data characteristics (e.g. spatial resolution), mapping approaches 78 (e.g. mapping unit) and forest definitions (e.g. tree cover thresholds). It has led to major

- doubts about the reliability of global forest area estimates, and to questions about thereal contribution made by forests to the global carbon cycle (*12*).
- 81

82 To address these uncertainties, we established a global initiative to undertake a Global 83 Dryland Assessment of forest. The geographical scope of this assessment is framed by 84 the delineation adopted by the United Nations Environment Programme World 85 Conservation Monitoring Centre (1), i.e. lands having an Aridity Index (AI) lower than 0.65. The AI is the ratio between average annual precipitation and total annual potential 86 87 evapotranspiration (16). The dryland domain is typically divided into four distinct 88 "zones" based on their AI: (i) the "hyperarid" zone (AI = <0.05), (ii) the "arid" zone 89 (AI = 0.05-0.2), (iii) the "semi-arid" zone (AI = 0.2-0.5) and (iv) the "dry subhumid" 90 zone (AI = 0.5-0.65). Using this definition, drylands cover 6,132 Mha, or 41.5% of the 91 Earth's land surface (1) (Fig. S1). Our study aims to determine accurately how much 92 forest and tree cover remains in dryland biomes.

93

94 Mapping forests in the drylands using satellite data is challenging, even with high 95 spatial resolution imagery (10-30 m). This is due to difficulties in (i) disentangling the 96 reflectance of trees, bare soil and the darkening effect of tree crown shadows in open 97 forests (17, 18), and (ii) detecting forest presenting a closed canopy with a low vegetative reflectance, such as Acacia or Eucalyptus species (18, 19). To overcome 98 99 these limitations, we took advantage of recent developments in cloud computing (20), 100 especially the suite of Google geospatial tools, which have greatly increased the 101 capacity to access and analyse large remote sensing databases of Very High spatial 102 Resolution (VHR) images (with a pixel width ≤ 1 m). VHR images allow scientists to 103 visually identify individual tree crowns in dry areas, e.g. of common genera such as *Adansonia* (baobab) in Africa (21) and *Acacia* in Australia (Figs. S2 and S3). Terrestrial
land coverage with VHR images is nearly complete (22), and this is the first study to
use them for global mapping purposes.

107

108 To determine the extent of forests and tree cover throughout the world's dryland 109 biomes, we assessed a large sample of 0.5 ha plots through visual interpretation of VHR 110 images available from Google Earth. We designed a stratified systematic sample with 111 higher sampling intensity from hyperarid to dry subhumid zones, leading to 213,795 112 sample plots (17; Fig. S4). To interpret the VHR images over such a large number of 113 plots we divided the world's dryland domain into 12 regions and employed a 114 participatory approach. Scientists and students in 15 organizations around the world 115 (Fig. S5) were trained to use a dedicated interpretation tool called Collect Earth (23) 116 with a common framework to assess the sample plots in which they had expertise.

117

118 Over 70 land attributes were assessed in each plot, but only forest and tree cover results 119 are reported here. Forest area and tree cover percentage were considered independently 120 to enable comparison with previous estimates. The tree cover percentage is assessed at 121 each plot irrespective of its land use type. Time series of vegetation indices for the 122 period 2000-2015 were computed from high temporal resolution satellite imagery 123 (MODIS and Landsat), and are used here to assist visual interpretation of VHR satellite 124 imagery (17; Fig. S2D). Trees were distinguished from shrubs by considering crown 125 shadows, which are related to vegetation height, and by using field-based photographs 126 available from the Web. Where information or knowledge was not sufficient for 127 distinguishing trees from shrubs, a tree crown diameter threshold of 3 m was applied.

128

Data quality was controlled through a semi-automated data cleansing procedure that
automatically identified potential inconsistent plots that were then manually reassessed.
Uncertainties were assessed by accounting for the sampling and interpretation errors,
the latter being assessed from 441 reference field plots (*16*).

133

134 Our results show that in 2015 there were $1,327 (\pm 98)$ Mha of dryland where tree canopy 135 cover percentage is over 10%, of which 777 Mha (57%) present a closed canopy (Table 136 1, Table S1), i.e. with a tree canopy cover over 40% (24). There are significant 137 differences between continents, e.g. half the total area with more than 10% tree cover 138 is located in Africa and Asia, and more than one third in North and South America 139 (Table 1; Figs S6-7). Of these 1,327 Mha, 1,079 (±38) Mha are considered as "forest" 140 according to the FAO definition (24): land spanning an area of more than 0.5 ha with a 141 tree cover over 10% that is not predominantly used for agriculture or urban land use, as 142 well as land on which tree cover is temporarily under 10% but is expected to recover 143 (Table S1, Fig. 1). Our estimates for the area with more than 10% tree canopy cover 144 and the area of forest differ by 271 Mha, or 23% (Fig. S8). This might help to explain 145 the 19% difference between recent estimates of forest "land use" area (3,890 Mha) (25) 146 and the area with a "land cover" presenting more than 10% tree canopy cover derived 147 from a global tree cover map (4,628 Mha) (13).

148

Our findings show that the total area of dryland forest is similar to the area of tropical moist forest, estimated at 1,156 Mha in 2000 (*15*). Its distribution is concentrated to the south of the Sahara desert, around the Mediterranean sea, and in southern Africa, central India, coastal Australia, western South America, northeast Brazil, northern Colombia

and Venezuela and in the northern belt of boreal forests in Canada and the RussianFederation (Fig. 1).

155

Almost two thirds of all dryland forests are closed canopy forests (Table 1, Table S1).
Open forests cover 355 Mha and are dominant in Africa and Oceania, where they
account for 52% and 74% of all dry forest, respectively. Of the total area of 1,079 Mha
of dryland forest, 523 Mha are located in the tropics, of which 203 Mha (37%) are open
forest and 320 Mha (63%) are closed forest (Supplementary Table 2).

161

162 When we compared our maps of forest and tree cover, based on +210,000 sample plots, 163 to recent maps based on coarser resolution satellite imagery (13, 14, 25, 26), we found 164 that the latter maps were missing significant areas of tree cover and forest in dryland 165 biomes (Table 2, 17, Figs. S9-11). Our estimate of 1,327 Mha for areas with over 10% 166 tree canopy cover is 427 Mha (47%) and 378 Mha (38%) higher than estimates derived 167 from the full drylands extracts of Hansen et al.'s 2000 map (13) and Sexton et al.'s 2010 168 map (14), respectively (16). These differences are of the same order as the total area of 169 tropical moist forest in Amazonia. The gaps tend to increase in regions with a high 170 proportion of open forest (Fig. S12), which illustrates the limitations of using medium-171 to-high resolution satellite images to identify low tree cover (27), and explains why the 172 gaps are particularly important in Africa and Oceania (Figs. S9-11). In Africa, for 173 example, we find 148 Mha (70%) more land with $\geq 10\%$ tree canopy cover than Hansen 174 et al., with the largest discrepancy observed in the Sahel and southern Africa (Fig. 2). 175 The differences for closed canopy forest (with> 40% tree cover) are even larger, as our 176 estimate for Africa is 151 Mha (Table 1), compared with only 18 Mha in Hansen et al. 177 and 2 Mha in Sexton et al. (Table S2, Fig. S11). We find even more tree cover and forest than the 2009 Globcover product (27) and the FAO-FRA global Remote Sensing
Survey 2010 (26), respectively (Table 2).

180

The global maps of Hansen et al. (2013) and Sexton et al. (2013) show some areas of $\geq 10\%$ tree canopy cover that are not apparent in our map, e.g. in NE Brazil and South-Sudan (Fig. 2, Figs. S10, S13). We suspect that these are caused by a 'greening effect' related to meadows or wetlands, i.e. which might present a spectral signature similar to forests and to which Landsat data are sensitive (*17*). 186

Our estimate is 40-47 % higher than previous estimates of the extent of forest in drylands. This potentially increases by 9% the global area with over 10% tree canopy cover (5,055 Mha instead of 4,628 Mha (*13*)) and by 11% the global area of forest (4,357 Mha instead of 3,890 Mha (*25*)).

191

192 Using numbers on the carbon pools of woody savannas (28), further research could use 193 our publicly available data to increase estimates of global forest carbon stocks by 15 to 194 158.3 GtC, or by 2 to 20 % (29), thereby helping to reduce uncertainty about the global 195 carbon budget (30). Our findings could also lead to the development of innovative 196 conservation and land restoration actions in dryland biomes, i.e. in regions with low 197 opportunity cost, to mitigate climate change, combat desertification, and support the 198 conservation of biodiversity and ecosystem services that underpin human livelihoods 199 (31).

200 **Table 1.** Areas in the world's drylands in 2015 of forest (as defined by FAO(24)) and

	Total area	Tree canopy cover ≥ 10%	Forest	Tree canopy cover ≥ 10 & < 40%	Open forest	Tree canopy cover ≥ 40%	Closed forest
Continent							
Africa	1961	364	286	213	151	151	135
Asia	1950	299	213	104	37	195	176
Europe	295	92	63	29	7	63	56
N America	694	238	204	77	49	161	155
Oceania	685	124	114	94	85	30	29
S America	546	208	197	33	26	175	171
Aridity zone							
Hyper-arid	978	13	3	9	2	4	1
Arid	1566	103	71	75	50	28	21
Semi-arid	2263	559	440	283	186	276	254
Dry sub-humid	1326	652	565	183	117	469	448
Drylands total	6132	1327	1079	550	355	777	724

201 land under different percentages of tree canopy cover (Mha).

202

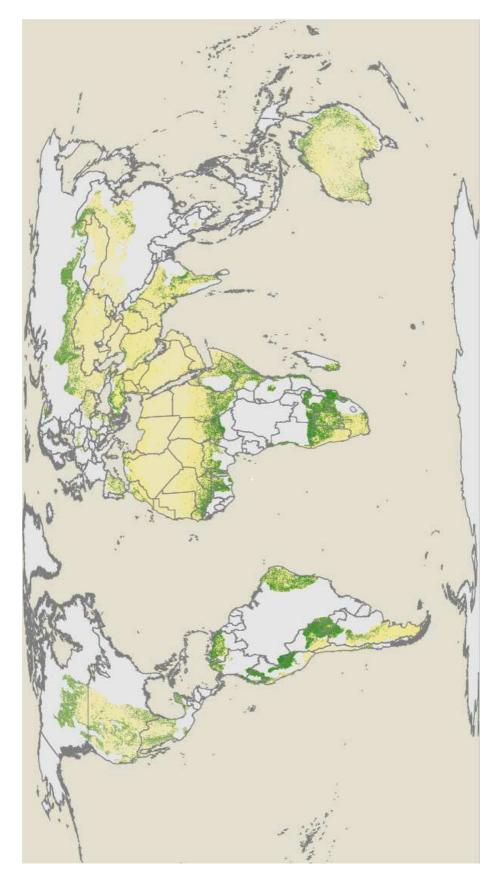
NB. Forest (column 3) is land with $\geq 10\%$ tree canopy cover that is not used for agriculture or settlement, or has <10% tree canopy but is regenerating; open forest (column 4) is forest with 10-39% tree canopy cover; closed forest is forest with $\geq 40\%$ tree canopy cover

203**Table 2.** Comparison of the estimate in this paper (Global Dryland Assessment) of204areas in the drylands in 2015 with forest and $\geq 10\%$ tree canopy cover (Table 1), with205other estimates based on satellite images and following the same definition of dryland

206 (Mha) (1).

Source	FAO RSS (2010) (25)	Globcover (2009) (<i>26</i>)	Hansen et al. (2013) (13)	Sexton et al. (2013) (14)			(2016)
Sensor	Landsat	MERIS	Landsat	Landsat	Ve	ry high re	solution
Method	sampling	wall-to-	wall-to-	wall-to-		S	ampling
		wall	wall	wall			
Year	2010	2008	2000	2010	2015	2015	2015
Forest	Yes	-	-	-	Yes	-	-
Tree cover	-	≥15%	≥10%	≥10%	-	≥20%	≥10%
Africa	67	83	216	114	286	253	364
Asia	43*	148	154	200	213 (97*)	242	299
Europe	22*	49	97	116	63 (26*)	78	92
N America	166	155	173	196	204	201	238
Oceania	29	28	55	55	114	71	124
S America	123	46	205	268	197	192	208
Total	450	509	900	949	1079 (917*)	1037	1327

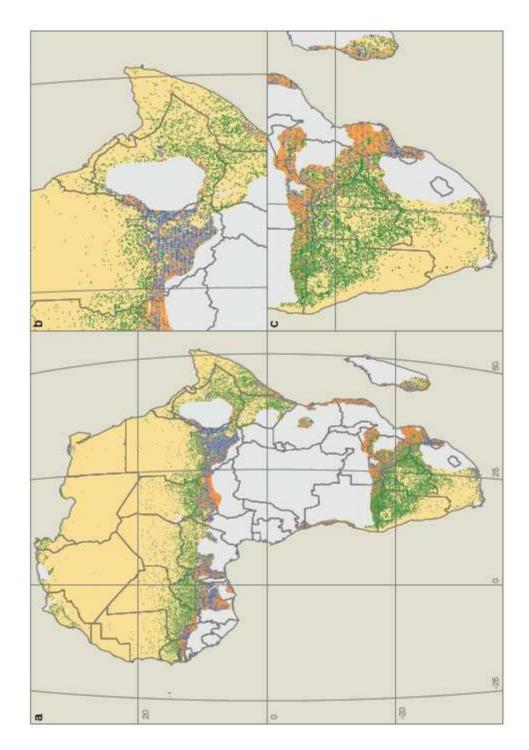
207 * Without Russian Federation





209 Figure 1. Forest distribution in drylands. Plots with forest are coloured in green,

and without forest in yellow.



211

Figure 2. Comparison of $\geq 10\%$ tree cover in Africa's drylands as mapped by the Global Drylands Assessment (GDA) and Hansen et al. (13). Green dots show plots are coloured green where the GDA reports $\geq 10\%$ tree cover but Hansen et al. reported a lower percentage; blue dots show plots where Hansen et al. reported $\geq 10\%$ tree cover but the GDA reports a lower percentage; and orange dots show plots where both assessments report $\geq 10\%$ tree cover. Figures 2b and 2c focus on two regions with large discrepancies between the maps.

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329

330 Author contributions

J.-F.B., D.Ma. and D.Mo. conceived and designed the paper. J.-F.B., D.Mo., A.G. and

332 R.C. wrote the paper, J.-F.B. and N.P. did the statistical analyses. C.P and S.R

333 coordinated the data cleansing procedure. B.S., A.L. and G.G. coordinated the field

data collection. N.B., A.G., D.Ma., D.Mo., C.P., B.S., E.M.A., K.A., A.A., F.A.,

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337 All authors assisted editing the manuscript.

338

339 Competing financial interests

340 The authors declare no competing financial interests.

342 Supplementary materials

- 343 Materials and Methods
- 344 Tables S1 to S3
- 345 Figs. S1 to S18
- 346 References (32-39)