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4	Human population growth offsets climate-driven increase in woody vegetation in sub-Saharan Africa
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26 Abstract

27 The rapidly growing human population in sub-Saharan Africa generates increasing demand for agricultural land 28 and forest products which presumably leads to deforestation. Conversely, a greening of African drylands has 29 been reported, but this has been difficult to associate with changes in woody vegetation. There is thus an incom-30 plete understanding of how woody vegetation responds to socio-economic and environmental change. Here we 31 used a passive microwave Earth Observation data set to document two different trends in woody cover land area 32 for 1992-2011: an 36% increase (6,870,000 km²), largely in drylands, and an 11% decrease (2,150,000 km²), 33 mostly in humid zones. Increases in woody cover were associated with low population growth and driven by 34 increases in CO₂ in the humid zones and by increases in precipitation in drylands, whereas decreases in woody 35 cover were associated with high population growth. The spatially distinct pattern of these opposing trends re-36 flects (1) the natural response of vegetation to precipitation and atmospheric CO_2 and (2) deforestation in humid 37 areas, minor in size but important for ecosystem services, such as biodiversity and carbon stocks. This nuanced 38 picture of changes in woody cover challenges widely held views of a general and ongoing reduction of the 39 woody vegetation in Africa.

40 Introduction

41 Africa's human population has increased from about 230 million in 1950 to over 1000 million in 2010 and is 42 expected to grow to as high as 5700 million by the end of the 21^{st} century¹. This growth has led to the expansion of agricultural land and the reduction of natural forests and other woody vegetation^{2,3,4}, affecting biodiversity and 43 carbon storage³. Severe droughts in recent decades have also had an adverse impact on humid and sub-humid 44 forested areas⁵. In contrast, studies of drylands have shown an increase in vegetation productivity over the last 45 30 years^{6,7,8}, also highlighting the importance of drylands for global carbon variability and as land CO₂ sink⁹. 46 Whether this increase in vegetation productivity is driven by the growth of woody vegetation and/or by an in-47 crease in productivity of herbaceous vegetation is not clear^{6,7,8}. This is because the scattered nature of woody 48 49 plants in drylands is very different from forests with closed canopies and challenging to detect with optical satellite imagery at regional to continental scales^{10,11}. Previous studies have used vegetation indices as proxies for net 50 51 primary productivity, but these indices measure the photosynthetically active part of the vegetation and most studies do not distinguish between woody and herbaceous vegetation^{12,13}. Furthermore, studies of deforestation 52 in humid areas traditionally report the presence or absence of forests³ and do not assess gradual changes in forest 53 54 biomass within existing forests (e.g., forest degradation). They are also based on temporal snapshots of satellite 55 imagery at a higher spatial resolution and only capture forests based on given definitions, e.g. tree height and canopy cover percentage^{3,14}, which substantially underestimate shrubs and scattered trees in drylands¹⁰. Conse-56 57 quently, little quantitative information is available about the state, rate, and drivers of change in the cover of 58 woody vegetation at the scale of the African continent. This information is crucial for ensuring that the design of 59 natural resource management in relation to deforestation and desertification is based on observations rather than 60 those based on narratives.

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62 Results

63 Africa's changing woody cover

64 We used a new passive microwave Earth Observation (EO) data set (Vegetation Optical Depth, VOD) that cap-

65 tures continuous changes in the coverage of canopies of all woody phanerophytes, regardless of size, in both drylands and humid areas¹⁵⁻¹⁷. We applied VOD as a proxy for annual woody cover and documented changes in 66 67 Africa's woody vegetation between 1992 and 2011, with a special focus on the changes in drylands and humid 68 areas (defined by the ratio between annual precipitation and potential evapotranspiration, Supplementary Fig. 69 1a). Woody vegetation changed significantly (linear regression, p < 0.05, n=20) during 1992–2011 in approxi-70 mately half of sub-Saharan Africa (47% of land areas). A majority (77%) of the significant trends were positive, 71 covering 36% of sub-Saharan Africa and representing an overall increase of 2.1 woody cover (%) (Fig. 1a). Most 72 (70%) of the significant positive changes were in drylands covering approximately 4 900 000 km² (overall change +2.9 woody cover (%)), mainly in the Sahel and southern Africa¹⁸⁻²⁰ (Fig. 2a). Positive trends are also 73 74 observed in the humid zones to a much smaller extent (2 100 000 km²), with an overall change of +0.8 woody 75 cover (%). Negative changes affected 11% of sub-Saharan Africa, of which 75% were in humid areas (approxi-76 mately 1 600 000 km² in humid zones and 530 000 km² in drylands). The decline in woody cover primarily af-77 fected areas that are also characterized by high carbon stocks (Supplementary Figs 2a, 2b), suggesting that areas 78 with the largest carbon sinks have been disturbed at the fastest rate. The classification of woody cover change 79 into bioclimate zones²¹ confirms the overall tendency with larger increases in drier zones (except extremely hot 80 xeric) and lower increases and decreases in moister zones (Fig. 2d).

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82 Drivers of woody cover changes

83 The positive changes in woody cover in Africa's drylands are significantly related to precipitation (Fig. 1a). In 84 contrast to herbaceous vegetation, woody plants can benefit from a higher variability and intensity of precipita-85 tion²², as in southern Africa and the Sahel (Supplementary Fig. 1c). The dependence on precipitation was corroborated with simulations of the vegetation using the dynamic vegetation model LPJ-GUESS²³, which simulat-86 87 ed an increase in woody biomass for 1992-2011, consistent with the satellite estimates of woody cover (Fig. 3). 88 The relative increase of both woody cover and biomass was largest in drylands, and factorial simulations of the 89 individual driving variables indicated that precipitation accounted for most of the simulated increase in woody 90 biomass in drylands such as the Sahel and southern Africa (Fig. 3, Supplementary Fig. 3). Increasing concentra-

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91 tions of atmospheric CO_2 was a minor contributor to these dryland trends yet was the main variable driving the 92 growth of woody vegetation in humid areas, enhancing primary production²⁴ (Fig. 3, Supplementary Figs. 2, 3). 93 The absolute increase in woody biomass was largest in humid areas (mean increase of 0.04 kg C m⁻² y⁻¹ near the 94 equator, Supplementary Fig. 2), coinciding with overall large stocks of woody biomass. Solar radiation, nitrogen 95 deposition and temperature had minor impacts on the changes in woody biomass (Supplementary Fig. 2).

96 This overall increase in woody vegetation driven by climate and CO_2 , however, was offset by anthropogenic 97 impacts, especially in humid areas. The increase in woody cover in the VOD analysis was thus most pronounced 98 in areas of low human population density and change (Fig. 4). Areas and countries with a higher population den-99 sity and growth (Fig. 1b, Supplementary Fig. 1d) had decreases in VOD-based woody cover (Figs. 1a, 4, Sup-100 plementary Fig. 4), offsetting the climate-driven increases in other parts of the humid zones (Figs. 2c, 3). This 101 separation in areas of high and low human pressure applied to both drylands and humid tropics. The average 102 trend, however, remained positive in drylands, even in areas with strong population growth, but was negative in 103 humid areas with strong population growth, regardless of the trends in precipitation and CO₂ (Fig. 2b, c). Popu-104 lations increased by an average of 40 persons km² over 20 years in areas where woody cover decreased suppos-105 edly due to agricultural expansion, logging, and other uses of woody products. In contrast, populations increased 106 by an average of only 6 persons km⁻² in areas where woody cover increased. Human population increase was 107 highest in moist and mesic bioclimate zones and woody cover changes were accordingly negative or low, where-108 as population growth was lower in xeric areas and woody cover increases were higher (Fig. 2d). At the continen-109 tal scale, a simultaneous autoregressive model (SAR) explained nearly half of the spatial pattern of changes in 110 woody cover in terms of changes in population and precipitation ($r^2=0.46$), with population being more im-111 portant than precipitation (standardized slopes of -0.27 and 0.08, respectively) (Supplementary Table 1).

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113 Discussion

114 The opposing trends in dry and humid zones have implications for our understanding of environmental change in 115 sub-Saharan Africa. While areas of high population growth, mostly in humid zones, on average experience a

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116 decrease in woody vegetation, areas with low population growth on average experience an increase in woody 117 vegetation, mainly driven by changes in precipitation and CO_2 concentrations. This latter increase is not captured 118 in official forest statistics, since much of it takes place outside of humid forests.

This implies that the 'problem' of woody cover loss - and thus carbon stocks decreases - in the humid forest zones is at least partly balanced by an increase in drylands. 'Bush encroachment' in savannas of southern Africa, however, has traditionally been considered an undesired effect^{14,25}. Since the VOD data used to estimate woody cover does not allow a direct estimation of carbon stocks, the exact balance between gains and losses in carbon cannot be directly assessed in this study. Further work combining field measurements, ecosystem modelling and new satellite-based passive microwave sensors is required to further understand these linkages. In humid areas, woody biomass may actually increase without any change in woody cover.

126 The close relationship between population growth and decreased woody cover suggests that agricultural expan-127 sion, urbanization and wood fuel harvest were the main causes of the decrease in woody cover, as also found in studies of tropical deforestation^{3,26}. The reduction in woody cover tends to primarily affect areas with high car-128 129 bon stocks and other studies suggest that these are also areas characterized by the highest biological diversity²⁷. 130 There is, however, no simple relation between losses and gains in woody cover and biodiversity. While diversity 131 and productivity of natural vegetation are generally positively correlated²⁸, this does not exclude the possibility 132 that great losses may be experienced in areas of deforestation, while only smaller gains are seen in drylands with 133 increasing woody cover.

Due to the impact on land surface albedo, woody cover changes in dryland areas may trigger climate feed-backs. Since the hypothesized existence of a 'biogeophysical feed-back'²⁹, many studies have attempted to model such effects^{30,31}, with some research claiming that man-made afforestation efforts would give rise to increased precipitation³². The extent of the observed increase in woody cover in African drylands may impact climate if the increase continues in the coming decades, and this altered feed-back should preferably be implemented in regional climate or Earth system models, with the observed increase in woody vegetation providing a test case for these models.

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142 Methods

143 VOD data and calibration to woody cover. We define woody cover as the percentage of a given area covered 144 by woody vegetation, including both leaf and woody components of woody plant canopies. The unit is woody 145 cover (%). The VOD data was retrieved from satellite passive microwave observations quantified as brightness temperature based on the NASA-VU Land Parameter Retrieval Model (LPRM)³³. Three passive microwave 146 sensors, i.e. the Special Sensor Microwave Imager, the Advanced Microwave Scanning Radiometer - Earth Ob-147 148 serving System, and the radiometer of WindSat are used to form the long-term data set by applying a trend-149 preserving cumulative distribution function matching without changing the inter-annual variations and long-term trends of the original retrievals^{34,35}. The merged long-term VOD data set was gridded at a 0.25° spatial resolution 150 and monthly interval from 1992 to 2011 and is consistent between different sensors³⁶. VOD is sensitive to the 151 152 total aboveground water content in both the photosynthetic (foliar) and non-photosynthetic (woody) components of the vegetation stratum^{15,37}. Soil moisture conditions are retrieved simultaneously with the VOD information in 153 154 LPRM and large variations in soil moisture can influence the accuracy of VOD, especially for dense rainforest regions³³. Thus VOD values exceeding 1.2 are suggested to be excluded in vegetation studies¹⁶. The VOD signal 155 has been separated from soil moisture and is used as a proxy for vegetation biomass globally³⁴. The VOD sea-156 157 sonal variation is a combined effect of the seasonal dynamics of both herbaceous (including crops) and woody vegetation¹⁵. We used the annual minimum VOD values as a proxy for woody vegetation cover to minimize the 158 influence of annual herbaceous vegetation¹⁰ and avoided values exceeding 1.2 (Supplementary Fig. 5). Areas 159 160 with perennial herbaceous vegetation may lead to an over-estimation of woody cover; however, the woody cover 161 in % is usually higher in these areas concealing the influence from the herbaceous plant understory. Also, VOD 162 data have been used to estimate forest change in South America by limiting the range of VOD values to 0.6-163 1.2^{16} . We did not restrict the VOD range to also include young trees and shrubs, which form an important part of the community of woody vegetation. Minimum VOD agrees well with a field data based map of woody cover 164 for Sahel $(r^2=0.80)^{10}$ (Supplementary Fig. 5). A global map calibrated with optical high spatial resolution images 165 and also assessing smaller trees produced similar results³⁸ and was thus used to transform the annual minimum 166 167 VOD to the unit woody cover (%) for further analyses ($r^2=0.85$, slope=0.86) (Supplementary Fig. 5). A third168 degree polynomial regression was used for the transformation. Woody cover <10% was predicted with an expo-169 nential regression to avoid underestimation of very low values. The VOD is insensitive to the effects of atmos-170 pheric and cloud contamination, ensuring reliable retrievals in cloudy regions e.g. central Africa.

171 Correlation between the trends in woody cover and changes in human population and precipitation. Precipitation data were derived from the Climate Research Unit (CRU) (data set version 3.23), which is globally 172 available for a 0.5° grid at monthly scale and is based on the upscaling of data from rain gauges³⁹. CRU precipi-173 174 tation data intrinsically includes some uncertainty, as the number of stations used for each grid cell varies con-175 siderably between cells and years. Even though it is the most widely used precipitation data set in dynamic vegetation modelling⁴⁰ and consistency with other data sets has been shown⁴¹, results have to be considered with cau-176 177 tion⁷. We have tested the blended GPCP data set, without significant changes of the results, still it has to be not-178 ed that a linear trend analysis on annually summed data includes uncertainties and simplification. We summed the monthly observations to obtain annual sums from 1992 to 2011 and resampled the data to 0.25° using a bicu-179 bic interpolation. Population data were acquired from Gridded Population of the World (GPW) v3⁴², which in-180 cludes estimates for 1990, 1995, 2000, 2005, and 2010, gridded with an output resolution of 2.5 arc-minutes, 181 182 resampled for this study to 0.25° (nearest neighbor). GPW population data were acquired from national statisti-183 cal offices and gridded based on the proportional method, which allocates population counts to grid cells based 184 on the proportion of each administrative areal unit that overlaps the cell. The gridded counts for existing census 185 years are then projected to the set of output years based on a simple model of population growth. The modeling 186 was thus not based on any additional layers of data, such as land cover, avoiding potential problems of endoge-187 neity between VOD and simulated population grids. A linear trend analysis was conducted for annual woody 188 cover and precipitation data, and the slope multiplied with the number of years to retrieve the absolute change 189 over time in the corresponding unit, facilitating the direct comparison with the human population data. We quan-190 tified the relationships between the changes in woody cover (estimated by VOD), population increase (GPW), and precipitation (CRU) by applying a simultaneous autoregressive model (SAR) (spatial error type⁴³) to the 191 192 three gridded data sets. The SAR model accounts for spatial autocorrelation and uses change in woody cover as 193 response and log(change in population) and change in precipitation as explanatory variables. The logarithm of the human population data was applied since the relation between woody cover changes and human population is non-linear at pixel scale, i.e. if a high number of population is reached (mostly in cities), the woody cover stops to decrease further. Standardized variables were used to enable model coefficients inter-comparison (standardized variable = (variable - mean) / standard deviation).

Fires frequently occur in most African ecosystems. However, at the spatial and temporal scale of our analysis, we do not expect changes in fire regimes as a major cause of changes in woody cover in itself but rather as a consequence of human induced deforestation and land use change⁴⁴.

Dynamic ecosystem model. The dynamic ecosystem model LPJ-GUESS²³ was applied to simulate changes in 201 woody-biomass carbon in natural vegetation for 1992-2011. LPJ-GUESS simulates the distribution of plant 202 203 functional types, and each type is represented by four pools of biomass carbon: leaves, roots, sapwood, and 204 heartwood. The latter two were added to represent the amounts of stem (wood) carbon. This variable is closely 205 related to the woody cover estimated by VOD, but especially in tropical forests, differences are expected, as VOD is not able to fully penetrate the tree crowns⁴⁵. Simulations were run for 1992-2011, applying monthly 206 207 climate data (temperature, precipitation, sunshine duration) from meteorological stations, gridded to $0.5^{\circ} \times 0.5^{\circ}$ resolution (CRU TS 3.21³⁹), monthly model-derived estimates for nitrogen deposition⁴⁶, and annual mean at-208 mospheric CO₂ concentrations^{47,48} based on ice-core data and atmospheric observations as forcing. Land use and 209 210 land use change were not accounted for in the simulations, which were only applied to quantify the changes in 211 natural vegetation. The simulations were preceded by a two-stage spinup: For the first stage, vegetation growth 212 starts from bare-ground conditions, using climatic data for 1901-1930, and CO_2 levels were kept constant at the 213 concentration for 1901. For the second stage, representing 1901-1991, the actual climate data, atmospheric CO₂ 214 concentration and N deposition were used.

In addition to a full simulation with the forcing as described above, five factorial simulations were performed to separate the impact of individual driving variables. Only one of the four parameters (temperature, precipitation, radiation, or CO₂) was applied using the transient data as described above, whereas the other three parameters used a climatology for 1992-2011, applying monthly means over this 20-year period for the climatic parameters and an annual mean for CO₂. In the fifth factorial simulation, similar to the transient CO₂ simulation above, the 9

changing CO_2 concentration was combined with a climatology for N deposition, to separate the impacts of atmospheric CO_2 and N deposition on the CO_2 fertilization. These simulations were applied to determine the impact of the individual driving variables on the simulated trend.

223 Data availability CRU precipitation data are available from the University of East Anglia 224 (http://www.cru.uea.ac.uk/). The global tree cover map is available from the Geospatial Information Authority of 225 Japan, Chiba University (http://www.iscgm.org/gm/ptc.html#use). VOD raster data are developed by Yi Liu, 226 University of New South Wales. Gridded population provided CIESIN maps are by 227 (http://sedac.ciesin.columbia.edu/). Humidity zones are available from http://www.grid.unep.ch/index.php. DGVM results are available from the corresponding author upon reasonable request. 228

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conducted the analyses with support by F.T., J.P., R.F. and J.P.. K.R. and M.B. drafted the manuscript with contributions by all authors.

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Author Information Reprints and permission information are available at <u>www.nature.com/reprints</u>. Correspondence and requests for materials should be addressed to M.B. (<u>martin.brandt@mailbox.org</u>). The authors declare no competing financial interests.

244 Figure legends:

Figure 1 | Changes in woody vegetation and human population over two decades. **a**, Significant trends of woody cover (VOD) for 1992-2011, separated by the presence or absence of a significant (p<0.05, n=12845) correlation with cumulative 2-year precipitation during this period. **b**, Changes in human populations for 1990-2010. The maps in (**a**) and (**b**) share a clear pattern, especially areas with a decrease in woody cover, and no relation to precipitation coincide with a high population pressure. **c**, SAR model of the changes between woody cover, precipitation (both 1992-2011), and population (1990-2010). The units are expressed as change in the corresponding unit over the period of analysis.

Figure 2 | **Changes in woody cover (VOD) in different humidity zones. a**, Areas with changes in woody cover (linear regression of change in woody cover for 1992-2011). Annual profiles of woody cover for areas of statistically significant changes in woody cover in **b**, drylands and **c**, the humid areas of sub-Saharan Africa (Supplementary Fig. 1a). Black lines characterize areas of high human population increase (>30 persons km⁻²) and grey lines areas of low human population increase (<10 persons km⁻²) for 1990 to 2010. **d**, Woody cover and human population changes are grouped according to bioclimatic zones²¹.

Figure 3 | Climatic drivers of changes in woody cover and biomass in sub-Saharan Africa. Relative trends (% of mean year⁻¹) for 1992-2011 in woody cover (estimated with VOD) and woody biomass (simulated with LPJ-GUESS) had similar patterns of change from north to south. The trends of woody biomass were mainly driven by CO₂ (humid areas) and precipitation (drylands) (Supplementary Figs. 2, 3).

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Figure 4 | Links between changes in woody cover and human population. Intervals of mean population density (1990-2010, Supplementary Fig. 1d) were used to group the changes in woody cover (VOD) associated with population increases and the number of pixels showing significant woody cover change. A Chi-squared test between woody cover and population change indicated the statistically significant dependency between the two variables.