

# Constraining a land cover map with satellite-based aboveground biomass estimates over Africa

Guillaume Marie<sup>1</sup>, B. Sébastiaan Luyssaert<sup>2</sup>, Cecile Dardel<sup>3</sup>, Thuy Le Toan<sup>4</sup>, Alexandre Bouvet<sup>4</sup>, Stéphane Mermoz<sup>4,6</sup>, Ludovic Villard<sup>4</sup>, Vladislav Bastrikov<sup>5</sup>, Philippe Peylin<sup>1</sup>.

<sup>1</sup>Laboratoire des Sciences du Climat et de l'Environnement (LSCE/IPSL), CEA-CNRS-UVSQ, Université Paris-Saclay, Gif-sur-Yvette, France

<sup>2</sup>Faculty of Science, Vrije Universiteit Amsterdam, Amsterdam, The Netherlands

<sup>3</sup>Laboratoire Géosciences Environnement, Paul Sabatier University Université, Toulouse III, Toulouse, France

<sup>4</sup>Centre d'Etudes Spatiales de la Biosphère (CESBIO), Toulouse, France

<sup>5</sup>Science Partners Partner, Paris, France

*Correspondence*<sup>6</sup>GlobEO, Toulouse, France

*Correspondance* to : Guillaume Marie

(Guillaume.Marie@uantwerpen.be)(Guillaume.Marie@uantwerpen.be)

**Abstract.** Most land surface models can either, depending on the simulation experiment, calculate the vegetation distribution and dynamics internally by making use of biogeographical principles or use vegetation maps to prescribe spatial and temporal changes in vegetation distribution. Irrespective of whether vegetation dynamics are simulated or prescribed, it is not practical to represent vegetation across the globe at the species level because of its daunting diversity. This issue can be circumvented by making use of 5 to 20 plant functional types (PFT) by assuming that all species within a single functional type show identical land-atmosphere interactions irrespective of their geographical location. In this study, we hypothesize that remote-sensing based assessments of above-ground biomass can be used to refine, discretizing, constrain the process in which real-world vegetation is discretized in PFT maps. Remotely sensed biomass estimates for Africa were used in a Bayesian framework to estimate the probability density distributions of woody, herbaceous, and bare soil fractions for the 15 land cover classes, according to the UN-LCCS typology, present in Africa. Subsequently, the 2,55<sup>th</sup> and 97,55<sup>th</sup> percentile of the probability density distributions were used to create 2,5% and 97,5% confidence interval PFT maps. Finally, the original and refined constrained PFT maps were used to drive biomass and albedo simulations with the ORCHIDEE model. This study demonstrates that remotely sensed biomass data can be used to better constrain PFT maps. Among the advantages share of dense forest PFTs but that additional information on bare soil fraction is required to constrain the share of herbaceous PFTs. Even though considerable uncertainties remain, using remotely sensed biomass data, were the reduced enhances the objectivity and reproducibility of the process by reducing the dependency on expert knowledge and the ability to report the confidence allows assessing and reporting the credible interval of the PFT maps. Applying this approach at the global scale, would increase confidence in the PFT maps underlying assessments of present day biomass stocks which could be used to benchmark future developments.

## 1 Introduction

Degradation, fires and deforestation of tropical forests are responsible for two thirds of the global net deforestation emissions (Houghton et al., 2012; Le Quéré et al., 2015; Friedlingstein et al., 2020). Although African tropical rainforests represent only one third of the global tropical rainforests (Lewis et al., 2009), they were responsible for almost all, i.e., 1,48 PgC in 2015 and 1,65 PgC in 2016, of the net carbon (C) emissions of pan-tropical regions, but substantial uncertainty is associated with these estimates, i.e., 1,15 for 2015 and 1,0 PgC for 2016, mainly driven by fire and land use changes (Palmer et al., 2019). The uncertainty of model estimates, such as mentioned above, broadly comes from three sources: (1) the vegetation distribution in the model, (2) the ability of the model to simulate biomass accumulation of undisturbed vegetation, and (3) the ability of the model to simulate natural and anthropogenic disturbances of the standing biomass. As this study will focus on improving the description of the vegetation distribution, the first question that needs to be answered is why vegetation distribution remains so uncertain?

a mis en forme : Police :12 pt, Français (France)

a mis en forme : Normal, Espace Après : 3 pt, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Automatique

a mis en forme : Couleur de police : Automatique

a mis en forme : Couleur de police : Automatique

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Automatique

a mis en forme : Couleur de police : Automatique

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

Most land surface models can either calculate the vegetation distribution internally by making use of biogeographical principles (Sitch et al., 2003; Krinner et al., 2005; Clark et al., 2011) or use vegetation maps to prescribe spatial and temporal changes in vegetation distribution. Where the first approach results in a description of the potential vegetation, the second approach is more suitable when actual vegetation is to be studied. Irrespective of whether potential or actual vegetation is studied, it is not practical to represent vegetation across the globe at the species level because there are already over 60,000 tree species (Beech et al., 2017), not to mention the diversity in herbs, forbs and mosses. Land surface models represent this daunting diversity by making use of 5 to 20 plant functional types (PFT) (Huete et al., 2016). The underlying assumption of plant functional types is that all species within a single functional type show identical land-atmosphere interactions irrespective of their geographical location (Huete et al., 2016; Bonan et al., 2002; Brovkin et al., 1997; Chapin et al., 1996). Discretizing real-world vegetation in PFTs is a first source of uncertainty.

When actual vegetation is the focus of a modelling study, the vegetation distribution will have to be prescribed. The construction of vegetation maps first requires real-world observations, typically through satellite-based remote sensing. Current remote sensing technology does not enable distinguishing individual tree species; hence, vegetation is observed as land cover types (Defourny, P., 2019), which group vegetation with similar sensory characteristics. Remote sensing observations themselves as well as classifying them in land cover types is ~~a~~are the second and third source of uncertainties (Hansen et al., 2013; Mitchard et al., 2014; Hurtt et al., 2004). Because the land surface models require the vegetation to be discretized in PFTs, which may differ between different land surface models, the land cover types will have to be remapped on PFT maps. The rules applied in remapping satellite-based land cover types in PFT maps is formalized in so-called “cross-walking tables” (CWT) (Poulter et al., 2011; Poulter et al., 2015), which are a ~~third~~<sup>fourth</sup> source of uncertainty (Hartley et al., 2017).

Although CWTs have been extensively used to create PFT maps (Wei et al., 2018; Wei et al., 2016; Poulter et al., 2011; Krinner et al., 2005), the process of associating land cover types with specific PFTs remains difficult to reproduce since several iterations of expert knowledge are required (Poulter et al., 2011; Poulter et al., 2015). Various land cover classifications exist, in particular the commonly used FAO (Food and Agriculture Organization) Land Cover Classification System (LCCS; Di Gregorio and Jansen, 2000). Most classes of the LCCS correspond to a mix of PFTs, which fractions are difficult to assess and likely variable across regions. For example, several classes are labeled labelled as a mosaic of vegetation types (i.e., “Mosaic of natural vegetation (tree, shrubs, herbs)”; see Table 2 in Poulter et al., 2015). Not surprisingly, efforts have been made to decrease the need of expert knowledge in favor of favour of more objective and reproducible approaches, e.g., classification rules based on a suite of improved and standard MODIS products (Wanxiao et al., 2008). Moreover, producing PFT maps from satellite-based land cover maps needs to become fully automated when the temporal frequency of satellite-based land cover and biomass maps increases, i.e., will increase thanks to the recent GEDI Lidar data (Dubayah et al., 2020) and/or future SAR missions like the NASA-ISRO Synthetic Aperture Radar (NiSAR) or BIOMASS missions (Le Toan et al., 2011; Quegan et al., 2019).

a mis en forme

a mis en forme

a mis en forme

a mis en forme : Couleur de police : Noir

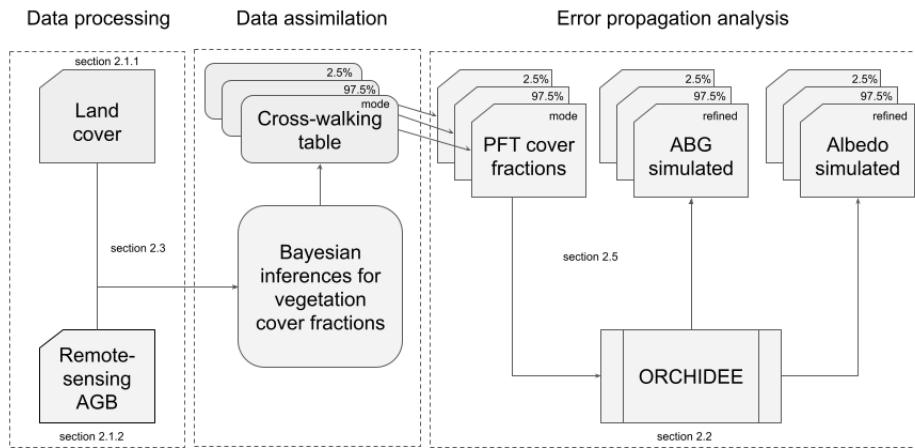
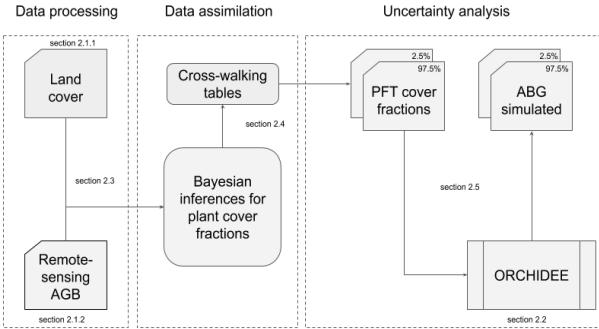
a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

In this study, we hypothesize that remote-sensing-based assessments of above ground biomass (AGB) can decrease the dependency on expert knowledge when setting up CWTs and as such contribute to the automation of the land cover class mapping into PFTs for land surface models. The main rationale is that the above-ground biomass content of an ecosystem provides information on the fraction of tree PFTs of that ecosystem. In this context, the objective of this study are: (1) construct a framework of data assimilation in which biomass remote sensing products can be routinely used to update an existing or create a new CWT, (2) refineconstrained a cross-walking table used to convert the ESA-CCI Global Land Cover map into a PFT map, and (3) propagate the confidentcredible interval from using a CWT in the production of PFT maps, to the simulation results of biomass and albedo maps derived from a land surface model. Such a framework will be applied and tested over Africa using the above ground biomass product derived by (Bouvet et al., 2018) for that continent with the ORCHIDEE land surface model (Krinner et al., 2005) and the version more specifically tag 2.0 revision 6592 close tag 2.2 used for the recent Climate Modelling Intercomparison Project - phase 6 (CMIP6) (Boucher et al., 2020).

## 2 Materials and methods

### 2.1 Overview

Cross-walking tables (CWT) (Poulter et al., 2015) are used to convert the 43 land cover types distinguished on the ESA-CCI land cover product into generic plant functional types (13 PFTs in (Poulter et al., 2015)) distinguished by large-scale land surface models such as the ORCHIDEE model (Krinner et al., 2005) used in this study. These generic PFTs are further grouped and/or divided to match each model-specific PFT classification, using additional grid-cell information to separate grassland and crop C3 versus C4 photosynthetic pathway (Still et al., 2003) and to split generic PFT according to bioclimatic zones (i.e., Koppen Geiger climate classification map) (see more details for the ORCHIDEE model in (Lurton et al., 2020)). In this study, we will createprovide a proof of concept by creating a new ORCHIDEE PFT map by combining information from the ESA-CCI land cover product and the AGB product for Africa (Bouvet et al., 2018) to estimate woody, herbaceous and bare soil cover fractions within each land cover type of the ESA-CCI product. Subsequently, the estimated cover fractions are used to refineconstrain the existing CWT and create a new ORCHIDEE PFT map applicable primarily for Africa (Fig. 1). Finally, the impact of using AGB maps to refineconstrain the PFT maps on the skill of the ORCHIDEE model to simulate the contemporary biomass and its spatial distribution over Africa is quantified. Note that the approach is tested over Africa but is generic enough to be applied everywhere.



**Figure 1:** Approach to assimilate the information held by aboveground biomass (AGB) maps into plant functional type (PFT) maps. Remote sensing AGB and land cover products are jointly assimilated to obtain cross-walking tables that can be used to make PFT maps. Owing to the uncertainty analysis in the data assimilation approach, an ensemble of cross-walking tables and PFT cover fraction maps can be produced. Subsequently, the land surface model ORCHIDEE can be run for different PFT maps to quantify the uncertainty from propagation of the uncertainty from remote sensing products into a model simulation.

**a mis en forme :** Normal, Espace Après : 10 pt, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

**a mis en forme :** Police : 9 pt, Gras, Non souligné

**a mis en forme :** Police : 9 pt, Gras

**a mis en forme :** Police : 9 pt, Gras

**a mis en forme :** Police : 9 pt, Gras

**a mis en forme :** Police : 9 pt, Gras

**a mis en forme :** Police : 9 pt, Gras

**a mis en forme :** Police : 9 pt, Gras, Couleur de police : Automatique

**a mis en forme :** Police : 9 pt, Gras, Couleur de police : Automatique

**a mis en forme :** Police : 9 pt, Gras

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

## 2.2 Dataset products

### 2.2.1 Land cover map

ESA's Climate Change Initiative for Land Cover (CCI-LC) produced consistent global LC maps at 300 m spatial resolution on an annual basis from 1992 to for the year 2015 (Defourny, P., 2019). Only one year (2015) has been used to estimate the new vegetation cover cross-walking table. The typology of CCI-LC maps follows the Land Cover Classification System (LCCS) developed by the United Nations (UN) Food and Agriculture Organization (FAO), to enhance compatibility with similar products such as GLC2000, and GlobCover 2005 and 2009. The UN-LCCS typology was designed as a hierarchical classification, which allows adjusting the thematic detail of the legend. The "level 1" legend, also called "global" legend, counts 22 classes and is globally consistent and thus suitable for global applications such as creating PFT maps for LSM-land surface models. The "level 2" or "regional" legend counts 43 classes which are not present all over the world and could be used in this study given its focus on a single continent, i.e., Africa (see section 2.2.3). In addition, the UN-LCCS partly overlaps with the PFTs used in climate models.

### 2.2.2 Aboveground biomass map

This study also makes use of a continental map of AGB of African savannas and woodlands for the year 2010 (Bouvet et al., 2018). The map has a 25 m resolution and is built from the 2010 L-band data of the Phased Array L-band Synthetic Aperture Radar (PALSAR) on the Advanced Land Observing Satellite (ALOS) satellite. Covering the African continent required about 180 data strips of which 91% were acquired between May and November 2010. The remaining 9% of the domain was filled with imagery from 2009 and 2008. The data have been processed by the Japan Aerospace Exploration Agency (JAXA) using the large-scale mosaicking algorithm described in Shimada and Ohtaki (2010), including ortho-rectification, slope correction and radiometric calibration between neighboring strips, and by Bouvet et al. 2018 (multi-image filtering), described in Bouvet et al., 2018.

The continental AGB map was derived as follows: (1) stratification into wet/dry season areas in order to account for seasonal effects in the relationship between PALSAR backscatter and AGB, (2) the development of a statistical model relating the PALSAR backscatter to observed AGB, (3) Bayesian inversion of the direct model, to obtain AGB and its confident credible interval for pixels where no observations are available, and (4) masking out non-vegetated areas using the ESA-CCI Land Cover dataset (but see section 2.1.1). The resulting AGB map was visually compared with existing AGB maps (Saatchi et al., 2011; Baccini et al., 2012; Avitabile et al., 2016) and cross-validated with AGB estimates obtained from field measurements and LiDAR datasets (Naidoo et al., 2015). Cross-validation revealed a good accuracy of the dataset, with an RMSD between 8 and 17 Mg·t·ha<sup>-1</sup>. For more details on the creation and evaluation of the AGB maps see Bouvet et al., 2018.

a mis en forme : Couleur de police : Noir
a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Couleur de police : Noir
a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

### 2.2.3 Pre-processing

One known limitation of the original AGB map (Bouvet et al., 2018) is the signal saturation and in some cases the decrease of the signal (Mermoz et al., 2015), occurring in L-band SAR for AGB values higher than 85 t·ha<sup>-1</sup>. In order to overcome this issue, a second AGB map was created, based on two other ancillary datasets: a map of tree cover (Hansen et al., 2013) and a map of tree height (Simard et al., 2011). Because of a coarser resolution from the tree height map (0,01°x0,01°, 100 ha) than the original AGB map (0,00025°x0,00025°, 0,0625 ha), the new biomass map has been rescaled to 0,01° resolution. The rescaling will also drastically reduce the noise produced by PALSAR measurement artefacts (personal communication). The above-ground biomass AGB was estimated by deriving an empirical relationship between biomass, available from airborne Lidar estimates, and the product of tree cover and tree height. The second version targets dense forest areas such as in the Congo basin (personal communication) and is used to adjust the AGB values at locations where signal saturation occurred (not published). Because of a coarser resolution from the tree height map (0,01°x0,01°, 100 ha) than the original AGB map (0,00025°x0,00025°, 0,0625 ha), the new biomass map has been rescaled to 0,01° resolution. The rescaling drastically reduced the noise produced by PALSAR measurement artefacts (personal communication Thuy Le Toan). The original AGB map was downscaled by an average resampling method, i.e., computing the weighted average of all contributing pixels. To do so, we used the Gdalwarp function from GDAL (GDAL/OGR). The map used in this study is a composite of the two versions of the biomass map by using the following rules:

- For broadleaved evergreen forests (UN-LCCS land cover type 50), flood forests (UN-LCCS 160), and closed broadleaved deciduous forests (UN-LCCS 61), the map based on tree cover and tree height was used because there is no AGB estimates in the map based on PALSAR.
- For broadleaved deciduous forests (UN-LCCS 60) the maximum between the two maps was used because its biomass ranged around the threshold of 85 t·ha<sup>-1</sup> and may create truncated distribution.
- For the other land cover types, which typically have a biomass well below 85t·ha<sup>-1</sup> the AGB value from the PALSAR map was used because it is considered more reliable than the statistical relationship between biomass, vegetation cover and vegetation height especially for the lower biomass.

Given the spatial domain of this study, only the 31 land cover types defined on the ESA CCI-LC map and present in Africa were retained. The complexity of the study was further reduced by removing all land types that cover less than 1,0% (304,158 km<sup>2</sup>) of the African surface or that contain less than 1% (i.e., 1,1 Gt) of the total AGB of Africa. Filtering retrained 15 out of the 31 land cover types including bare land. These 15 land cover types (Table 1) represent 96% of the surface of Africa and 98% of its AGB.

One additional issue had to be dealt with: the spatial resolution of the land cover map (9 ha) largely differed from the resolution of the AGB map (0,01°x0,01°, 100 ha). As a consequence Therefore, each observational point on the AGB map is represented by 11,111x11 pixels on the land cover map. In order to To simplify the overall data assimilation methodology (see section

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3,13", Centré + 6,27", Droite

3.2), we chose to use only AGB pixels (100 ha) which have a unique land cover type (i.e., pure pixels, in terms of LCC-land cover type). To this aim, the variety of land cover types across the 11,111x11 pixels within each AGB pixel (i.e., the number of LCC present,  $Vlct$ ) was calculated and only pixels where  $Vlct=1$  were retained. Although this criterion resulted in discarding 99% of the pixels, each of the 15 land cover types considered could be represented by at least 2000 pixels. In order to remove outlier pixels, we choose to pick up the 2000 pixels between strictly below the range of biomass value representing the 95% confidence interval/97.5<sup>th</sup> percentile of each LCT biomass distribution (see figure 2).

## 2.3 Data assimilation

### 2.3.1 Linking land cover fractions and AGB

A linear model was used to relate the satellite-based AGB of a 100-ha pixel to the cover fraction of the satellite-based vegetation types present at the same location. This relationship can be written as:

$$B_p = \sum_{i=1}^{nV} F_{p,i} \cdot Bref_i \quad (1)$$

$$B_p = \sum_{i=1}^{nV} F_{p,i} \cdot Bref_i \quad (1)$$

where  $B_p$  is the AGB at a given pixel  $p$ ,  $F_{p,i}$  is the cover fraction of the vegetation type  $i$  (i.e., the generic plant functional type (PFT) used for land surface models, see section 2.1 - overview),  $Bref_i$  is the reference AGB for the vegetation type  $i$  and  $nV$  is the number of vegetation types (i.e., number of PFTs) present in the pixel  $p$ . Given the number of unknowns ( $nV$  being usually above 1), equation 1 has many solutions; many of which have no biological meaning. The equifinality of this model can be reduced by arguing that the large difference in biomass between woody, herbaceous and non-vegetated ecosystems combined by their respective cover fraction explains the majority most of the biomass at pixel level. Following this assumption, equation 1 can be simplified as:

$$B_p = F_{p,w} \cdot Bref_w + (1 - F_{p,w} - F_{p,h}) \cdot Bref_h \quad (2.1)$$

$$B_p = F_{p,w} \cdot Bref_w + F_{p,h} \cdot Bref_h \quad (2.2)$$

$$B_p = F_{p,w} \cdot Bref_w + (1 + F_{p,h} - F_{p,b}) \cdot Bref_h \quad (2)$$

where  $F_{p,w}$ ,  $F_{p,h}$  and  $F_{p,b}$  are the fractions cover for woody vegetation (i.e., woody PFTs), herbaceous vegetation (i.e., grassland and cropland) and non-vegetated areas, respectively.  $Bref_w$  and  $Bref_h$  are the reference AGB of woody and

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

herbaceous vegetation, respectively. Equation 2.1 is constrained by equation 2.2 (i.e., the total area coverage of each pixel), hence,  $F_{p,b}$  in equation 2.1 can be substituted by equation 2.2 according to  $F_{p,w} + F_{p,h} + F_{p,b} = 1$  to obtain:

$$F_{p,w} + F_{p,h} + F_{p,b} = 1 \quad (3)$$

$$B_p = F_{p,w} \cdot Bref_w + F_{p,h} \cdot Bref_h \quad (3)$$

Although the model formalized in equation 3 no longer details which vegetation types i (i.e., PFTs) are present on each pixel p, it still has four unknowns and can, therefore, not be solved analytically. Nevertheless, a statistical solution is within reach if  $F_{p,w}$ ,  $F_{p,h}$ ,  $Bref_w$  and  $Bref_h$  are estimated from a population of AGB observations containing a number of several independent repetitions that largely exceeds the number of unknowns. In this study, over 2000 repetitions were available for each of the 15 land cover types that were retained following filtering (section 2.2.3). The statistical solution will thus consist of four mean parameter values (i.e.,  $F_{p,w}$ ,  $F_{p,h}$ ,  $Bref_w$  and  $Bref_h$ ) for each of these 15 land cover types.

As described in section 2.2.3, the selection of homogeneous AGB pixels, i.e., which have a unique land cover class across the 11–11 (i.e., 121) underlying land cover sub-pixels allow us to rewrite the equation 3 as follow:

$$\bar{Bp_p}$$

$$\bar{Bp_p} = F_{lc,w} \cdot Bref_{lc,w} + F_{lc,h} \cdot Bref_{lc,h} \quad (4)$$

where  $\bar{Bp_p}$  now  $Bp_p$  is the average AGB of a specific land cover type lc and  $F_{lc,w}$ ,  $F_{lc,h}$ ,  $Bref_{lc,w}$ ,  $Bref_{lc,h}$  are the unknowns. The unknown parameters of the regression model (eq. 4) were estimated by using a Bayesian inference method. This approach has been chosen because it helps to synthesize various sources of information as well as to propagate confident interval credible intervals in the result of our land surface model (Ellison 2004). Bayesian inference requires, however, setting prior probability distributions for each of the unknowns, i.e., the biomasses and land cover fractions for each of the 16 land cover types. Given these prior probability distributions, Bayesian inference retrieves the posterior probability distribution for each of the unknown parameters.

### 2.3.2 Prior value distributions for $Bref_{lc,w}$ , $Bref_{lc,h}$ and $Bp_p$

The pure AGB pixels were stratified according to their land cover type and for each land cover type the information contained in the distribution of the satellite-based AGB served to estimate the mean and standard deviation of the prior values of  $Bref_{lc}$ . In order to To avoid negative Bref values we used a normal truncated distribution with  $0 < a, b < +\infty$  where (a, b) are the truncated range:

$$Bref_{lc,w} \sim N(\mu_{lc,w}, \sigma_{lc,w}, a, b)$$

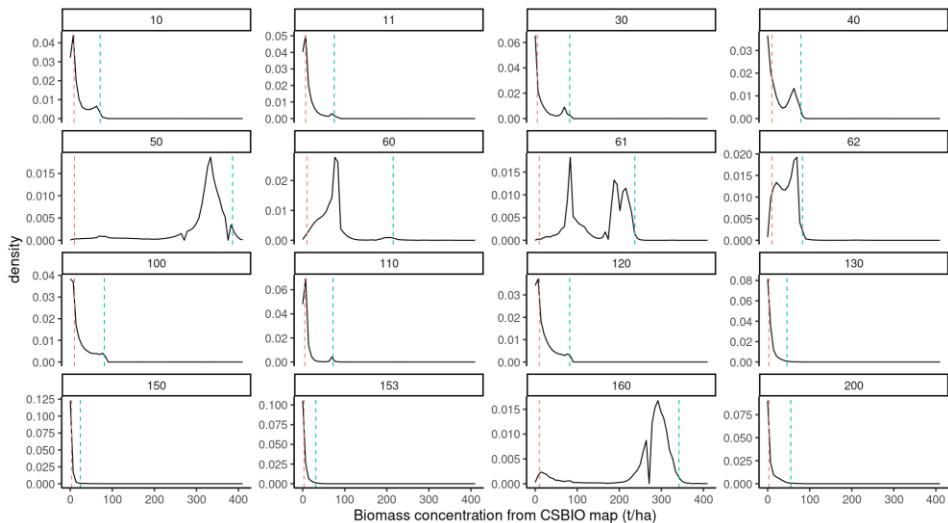
$$(\mu_{lc,w}, \sigma_{lc,w}, a, b)$$

where,  $\mu_{lc,w}$  is calculated as follow:

$$\mu_{lc,w} = X^{th} per. (Bp_{lc}) \quad (5)$$

$$\mu_{lc,w} = X^{th} per. (Bp_{lc}) \quad (6)$$

Where  $Bp_{lc}$  is a vector containing  $Bp_p$  values that belong to the land cover type lc and  $X^{th} per$  denotes the 97,5<sup>th</sup> percentile for the woody cover types. This choice assumes that with the highest 97,5<sup>th</sup> percentile we select the AGB value of a pixel covered only by woody vegetation (i.e., woody PFT) for the selected land cover type. In contrast to using a few in-situ observations to define  $\mu_{lc,w}$ , such an approach offers the advantage to rely on a large ensemble of satellite-derived AGB observations and to be coherent with the following optimization.



**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Police : Cambria Math, Couleur de police : Noir

**a mis en forme :** Gauche

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

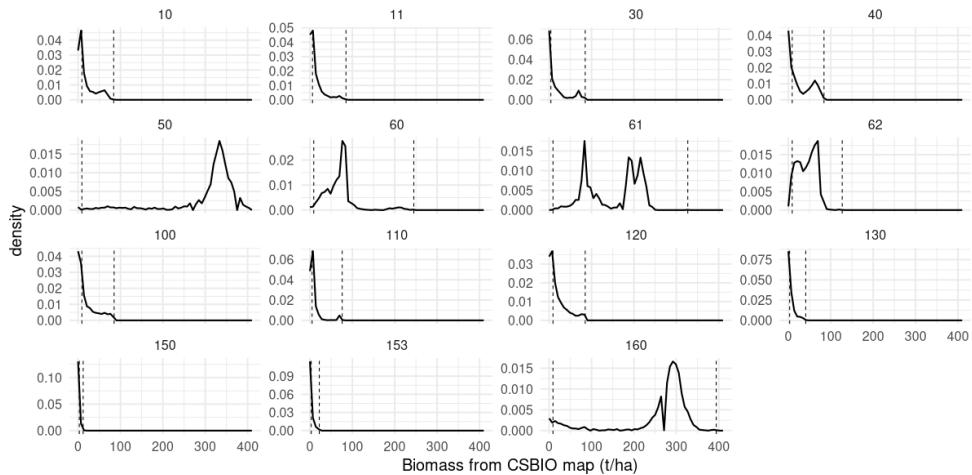


Figure 2: shows the Probability density distribution of the pure land cover pixel for biomass concentration  $B_{p,i}$  for 15 selected land cover types + LCT 200. The blue dashed line represents the 97.5<sup>th</sup> percentile used as the prior estimate for the reference biomass concentration for trees  $B_{ref,t}$ . The red dashed line represents the 50<sup>th</sup> percentile also used as the prior estimate for the reference biomass for herbaceous cover.

Without any information about the variability of  $B_{ref,w}$ , we choose to represent  $\sigma_{lc,w}$  as:

$$\sigma_{lc,w} = \frac{\mu_{lc,w} \cdot 0.15}{4}$$

Compared to  $B_{ref,w}$ ,  $B_{ref,c}$

$$\sigma_{lc,w} = \mu_{lc,w} \cdot 0.0375$$

(7)

a mis en forme : Police :9 pt, Gras, Non souligné, Couleur de police : Noir

a mis en forme : Police :9 pt, Gras, Couleur de police : Noir

a mis en forme : Normal, Espace Après : 10 pt, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Police :9 pt, Gras

a mis en forme : Police :9 pt, Gras, Couleur de police : Noir

a mis en forme : Police :9 pt, Gras, Couleur de police : Noir

a mis en forme : Police :9 pt, Gras, Couleur de police : Noir

a mis en forme : Police :9 pt, Gras, Couleur de police : Noir

a mis en forme : Police :9 pt, Gras, Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

Where 0.0375 accounts for a 30% uncertainty encompassed between the interquartile range of the normally distributed  $Bref_{lc,w}$ . Compared to  $Bref_{lc,w}$ ,  $Bref_{lc,h}$  is more difficult to assess from the satellite-derived data because it often shows bimodal distributions which may stem from biomass degradation or the presence of shrubs which biomass better resembles that of a grassland than a woody ecosystem (Fig. 2). We found that while the 2,5<sup>th</sup> percentile is representing the lowest biomass for herbaceous ecosystem, the 50<sup>th</sup> percentile seems to better describe  $Bref_{lc,h}$  following the equation (6). Without any Having no information about the variability of  $Bref_{lc,h}$ , we choose to represent  $b_{lc,h}$  as in followed equation (7).

Finally,  $Bp_p$  Finally,  $Bp_p$  which was the 97.5<sup>th</sup> for woody cover types or the 50<sup>th</sup> percentile for herbaceous cover types, comes with a measurement uncertainty that was thought to follow a normal truncated distribution with  $0 < a, b < Bref_{lc,w}$ , where (a, b) are the truncated range. Given that this measurement uncertainty is not known at the pixel level, an uninformative prior was set for the standard deviation  $\sigma b_{lc}$  which can varies between 0 and 200 t/ha. We deliberately took a large uncertainty to cover the fact a large observation that considerable uncertainty exists in the satellite-based biomass estimation method based on satellite data estimates (bouvet et al., 2018);

$$Bp_p \sim N(\mu, \sigma^2_{te}, a, b) \text{ with } \sigma b_{lc} \sim U(0, 200) \text{ and } \mu = Bp_p \quad (8)$$

### 2.3.3 Prior value distributions for $F_{lc,w}$ , $F_{lc,b}$ and $F_{lc,h}$

$F_{lc,w}$ ,  $F_{lc,b}$  and  $F_{lc,h}$  were defined as fractions of respectively woody vegetation and bare soil and herbaceous vegetation within a given land cover type, their values thus range between zero and one and their sum is equal to 1. For this reason, a Dirichlet distribution was used to describe the probability distribution of the woody, bare soil and herbaceous cover fractions. OpenBUGS (ref) do not allow using Dirichlet distribution as a stochastic node. To overcome this constraint, ones can use independent beta distribution. Let  $q_{lc,i}$ ,  $i = 1, \dots, K - 1$  a series of independently distributed  $Be(\alpha_i, \beta_i)$  distributions where  $K$  represent the number vegetation fraction. Let us define recursively from them our cover fractions:

$$(F_{lc,w}, F_{lc,b}, F_{lc,h}) \sim Di(\theta_{lc,t}, \theta_{lc,b}, \theta_{lc,h}) \quad (9)$$

OpenBUGS (Thomas, 2010), the software that was used in this study, cannot use a Dirichlet distribution as a stochastic node. This constraint can be overcome by making the cover fractions dependent on each other:

$$F_{lc,w} = q_{lc,1} \quad (10)$$

$$F_{lc,b} = q_{lc,2} \cdot (1 - q_{lc,1}) \quad (11)$$

$$F_{lc,h} = (1 - q_{lc,1}) \cdot (1 - q_{lc,2}) \quad (12)$$

Where i represent w (woody), b (bare soil) or h (herbaceous) described in the equation (5). Then a diriclet distribution can be reconstruct from :

Let  $q_{lc,i}$  with  $i = 1, \dots, K - 1$  and  $K$  the number of fractions, be a series of independent beta distributions,  $Be(\alpha_{lc,i}, \beta_{lc,i})$ .

$$(F_{lc,w}, F_{lc,b}, F_{lc,h}) \sim Di(\theta_{lc,t}, \theta_{lc,b}, \theta_{lc,h})$$

with:

$$q_{lc,i} \sim Be(\alpha_{lc,i}, \beta_{lc,i}) \quad (9)$$

(13)

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Normal, Gauche

a mis en forme : Police :10 pt, Couleur de police : Automatique

The parameters of the beta distribution of the cover fraction of bare soil, woody vegetation and herbaceous vegetation (eq. 9) can then be estimated as follows:

$$\alpha_{lc,i} = \theta_{lc,i} \cdot (\omega_{lc,i} - 2) + 1 \quad (10)$$

(14)

$$\beta_{lc,i} = \sum_{u=i+1}^K \alpha_{lc,u} \quad (11)$$

$$\omega_{lc,i} \sim U(0,1000) \quad (15)$$

where  $\theta_{lc,i}$  is which represents the fraction of each land cover type taken from expert knowledge used to define the so-called cross walking table (CWT) and taken from a recent update of the CWT (see ORCHIDAS, with an expert-based update of the original CWT described in Poulter et al., 2015).  $\omega_{lc,i}$  was described by an uninformative uniform distribution and represents thus reflects the relatively low trust we have in the current CWT. The dependencies between the beta distributions comes from  $\beta_{lc,i}$  that is estimated as:

$$\beta_{lc,i} = \sum_{u=i+1}^K \alpha_{lc,u} \quad (16)$$

a mis en forme : Police :Cambria Math

a mis en forme : Normal

a mis en forme : Police :Cambria Math

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

$$F_{lc,b} = U(0,1000) \quad (12)$$

## 2.4 Confident interval propagation

### 2.4.1 Propagating the ~~confident~~credible interval from the CWT into the PFT map

The posterior estimates of the cover fractions ( $F_{lc,w}$ ,  $F_{lc,b}$ ,  $F_{lc,h}$ ) will be directly used to make up a new cross-walking table. The posterior estimates of the cover fractions values are then used to recalculate woody and herbaceous fraction of each generic PFT of the CWT. In other words, we keep the original split of the different woody PFT defined in the prior CWT and only rescale the total woody fraction to  $F_{lc,w}$ . Then we rescale the bare soil fraction based on  $F_{lc,b}$  to finally rescale short vegetation PFTs (grass and crop) but using  $(1 - F_{lc,w} - F_{lc,h})$ .

Given that these posterior estimates come with a probability distribution, a probability distribution of the CWT could be made. In this study, the 2,5 and 97,5 percentiles and the mode, i.e., the most common value, of the posterior estimates were used to create three cross-walking tables that were then applied on the ESA-CCI-LC product to create two PFT maps that represent the 95% interval confidence of the ESA-CC-LC product and one PFT map which will representrepresents the one that is commonly used in an ORCHIDEE simulation, the AGB product, and the processing chain described in sections 2.2 and 2.3.

The impact of the various PFT map was quantified for simulated above ground biomass and simulated surface albedo by running three simulations that only differed by the PFT map used to initialize the ORCHIDEE land surface model.

In the study, the uncertainty propagation index aimed to identify the ecoregions where the AGB and surface albedo estimates are most sensitive to uncertainties from the PFT map. This sensitivity was calculated as:

$$S_{eco,b} = \frac{ABS(X^{97,5} - X^{2,5})}{ABS(F_{eco,b}^{97,5} - F_{eco,b}^{2,5})} \times 100 \quad (17)$$

Where  $X$  stands for AGB (t/ha) or surface albedo (unitless),  $S_{eco,b}$  is expressed in the unit of  $X$  for a 1% change in bare soil fraction.

### 2.4.2 Description of the ORCHIDEE land surface model

ORCHIDEE (Krinner et al., 2005; PeylinBoucher et al. *in prep.* 2020) is the land surface model of the IPSL (Institut Pierre Simon Laplace) Earth system model. As it is a land surface model, ORCHIDEEHenceHence, by conception, it can be coupled to a global circulation model. In a coupled setup, the atmospheric conditions affect the land surface and the land surface, in turn, affects the atmospheric conditions. However, when a study focuses just on changes in the land surface ORCHIDEE rather than on the interaction with the atmosphere, it also can be run as a stand-alone land surface model. The stand-alone configuration receives atmospheric conditions such as temperature, humidity, and wind, to mention a few, from the so-called meteorological forcing. The resolution of the meteorological forcing determines the spatial resolution which can cover range

from and can cover any area ranging from the global domain to a single grid point to the entire globe. ORCHIDEE uses nested time steps: half-hourly for, e.g., photosynthesis and energy budget, daily, e.g., net primary production, and annual, e.g., vegetation dynamics.

Although ORCHIDEE does not enforce a spatial or temporal resolution, the model does use a spatial grid and equidistant time steps. The spatial resolution is an implicit user setting that is determined by the resolution of the meteorological data. ORCHIDEE can run on any temporal resolution; however, this apparent flexibility is restricted as the processes are formalized at given time steps: half-hourly (i.e., photosynthesis and energy budget), daily (i.e., net primary production), and annual (i.e., vegetation dynamics). Hence, meaningful simulations have a temporal resolution of 1 min to 1 h for the energy balance, water balance, and photosynthesis calculations. In the land-only configuration used in this study, the default time step for these processes is 30 minutes.

When an application requires the land surface to be characterised by its In this study the model was run with 15 PFTs, where the additional PFTs represented tropical and boreal C3 grasslands which both belong to the meta-class of C3 grasslands. When an application requires the actual vegetation, the vegetation will have to be prescribed by annual land cover maps. These maps have to follow specific rules for the LSM land surface models to be able to read them. In the case of ORCHIDEE the share of each of the 15 possible plant functional types (PFTs) needs to range between 0 and 1 and be specified for each pixel. When satellite-based land cover maps are used as the basis for an ORCHIDEE-specific PFT map, the satellite-based land cover classification will need to be converted to match the ORCHIDEE specifications. As mentioned already above, this involves two steps: i) the derivation of generic PFTs from the satellite land cover classes (in our case the ESA-CCI-LC product) through the CWT discussed in this paper and ii) the final mapping of the generic PFTs into the 15 ORCHIDEE-specific PFTs using additional information on the bioclimatic zones and the partition of grassland/crops into C3 versus C4 photosynthetic pathway (Lurton et al., 2020).

In this study, AGB was defined as the sum of leaf biomass, fruit biomass, aboveground sapwood biomass, and aboveground heartwood biomass which are default output variables of ORCHIDEE. Surface albedo was defined as the albedo in the visible wavelengths and is a default output variable of ORCHIDEE.

**Table 1: Description of the 15 plant functional types (PFT) used in ORCHIDEE to represent global vegetation.**

PFT	Climate	Vegetation type	Phenology class
-----	---------	-----------------	-----------------

1	global	NA	Bare soil
2	Tropical	Woody	Broadleaf evergreen
3	Tropical	Woody	Broadleaf deciduous
4	Temperate	Woody	Needleleaf Evergreen
5	Temperate	Woody	Broadleaf Evergreen
6	Temperate	Woody	Broadleaf <u>Summergreen</u> <u>Summer green</u>
7	Boreal	Woody	Needleleaf Evergreen
8	Boreal	Woody	Broadleaf <u>Summergreen</u> <u>Summer green</u>
9	Boreal	Woody	Needleleaf Deciduous
10	Temperate	Herbaceous	Natural (C3)
11	global	Herbaceous	Natural (C4)
12	global	Herbaceous	Managed (C3)
13	global	Herbaceous	Managed (C4)
14	Tropical	Herbaceous	Natural (C3)
15	Boreal	Herbaceous	Natural (C3)

#### 2.4.3 Experimental setup

ORCHIDEE ~~tagstag~~ 2.0 (rev 6592) was used to run tree simulations that only differed by the PFT map used. Following a 340-~~yearlong spinup to initialize~~ ~~initialize~~ the carbon pools in the model. Each simulation consisted of a 110 years long simulation between 1901 to 2010 with the CRU-NCEP v8 climate reconstruction (Viovy, 2017) that matched the simulation years. CO2 concentration was fixed to 299,16 ppm ~~that and thus corresponds to the 2010 concentration~~ the CRU-NCEP/v8 was used as the climate forcing.

Finally, differences in simulated above ground biomass and surface albedo between the 2.5% and 97.5% were compared to each other as well as to the satellite-based AGB map [Bouvet et al 2018] and satellite-based albedo (REF). Note that in this study, the AGB by Bouvet et al (2018) was only used to constrain the cover fractions used by ORCHIDEE. Hence, the model-based estimate of biomass per unit area is independent from the AGB map as none of the information contained in that map was used in the ORCHIDEE model to simulate biomass per unit area.

#### 2.4.4 Ecoregions

In this study, we choose to represent results related to the LSM land surface model simulation were presented by subdividing the African continent into ecologically homogeneous regions. We choose to follow the rules, so-called ecoregions, as defined in the works of Olson et al., (2001).

### 3 Results

#### 3.1 Prior and posterior distributions estimates

##### 3.1.1 -Vegetation cover fraction: prior and reference biomass distributions

Prior distributions for the cover fractions and reference biomasses were determined for all 15 land cover classes separately, nevertheless, four broadly different groups could be distinguished: (1) The 97,5<sup>th</sup> percentile of biomass distribution for each land cover belonging in the first group was so high, i.e., from 245 to 416 t·ha<sup>-1</sup>, that the land cover types in this group must come with correspond to a substantial tree cover, i.e., a woody cover fractions of 0,58 to 0,75. Examples of this group are land cover types UN-LCCS 50, 61, and 160 (tree cover broadleaf types, in Table 2). (2) Contrary to the first group, the 97,5<sup>th</sup> percentile of biomass distribution for each land cover type of the second group is so low, i.e., from <12 to 42 t·ha<sup>-1</sup>, that these land cover types must be dominated by grasses or bare soil, i.e., a woody cover fraction of 0,1 or less and a substantial bare soil cover fraction i.e. from 0,01 up to 0,71. Examples of this group are UN-LCCS 130, 150 and 153 (grassland and sparse vegetation, in Table 2). (3) The biomass of the third group falls in between these extremes representing mosaic land cover types like the UN-LCCS 10, 11, 30, 40, 100, 110 and 120. (mosaic landscape in Table 2). When taken over the African continent, the biomass distribution of these land cover types shows bimodal biomass distributions indicating considerable variability within these land cover types (Fig. 2). (4) The bimodal biomass distribution of the fourth group is backed by a rather high woody reference biomass associated with a low woody cover fraction which may represent an ecosystem highly disturbed ecosystem i.e. sylviculturally either silvicultural practice and/or a fire regime. UN-LCCS 60, 62 fall into this group which represents the woodland to dry savanna continuum.

##### 3.1.2 -Vegetation cover fraction: posterior distributions

Owing to the Bayesian approach, the woody and herbaceous fraction within each land cover type is no longer deterministic (as was the case with the previous generation of cross-walking table such as in Poulter et al., 2015) but now comes with a distribution. This distribution is the outcome of propagating the confident credible interval on the retrieved parameters obtained from the Bayesian approach into the final product, i.e., the PFT cover fraction map. The 95% confidence credible interval was studied by comparing the 2,5 and 97,5 percentilepercentiles of the distribution of woody, herbaceous and bare soil fractions ( $F_{lc,w}$ ,  $F_{lc,h}$ ,  $F_{lc,b}$ ).

The mean change in forest cover fraction between the 2.5 and 97.5 percentilepercentiles of the distribution of refinedconstrained PFT maps over Africa was  $1.6\% \pm 2.6\%$ . At the ecoregion scale, (when averaging the cover fraction over the ecoregion), the largest confident interval uncertainty in forest cover fraction was found in the Congo basin with an average of  $-6.3\% \pm 0.5\%$  for the six concerned ecoregions i.e. ecoregions where LCT 50 is dominant (Fig 3). These results suggest that biomass map from CESBIO helps to reduce the uncertainty around the forest cover fraction. 3A.

In the contrary, the The 95% uncertainty interval for baresoilbare soil cover fraction is  $12.4\% \pm 13 \pm 8.9\%$ , mainly due to the large uncertainty of the cropland and mosaic croplant (LCTcropland (UN-LCCS, 10, 11, 30,40)). In ecoregions where in the concerned ecoregionthese LCTs are dominant, this credible interval riseincreases to  $24.6\% \pm 6.6$ . Dense $\pm 7\%$  (Fig. 3B). Moreover, dense forest land cover type i.e., LCT 50, 160 also experience a largecome with  $15\% \pm 4\%$  uncertainty from the baresoil covertonin their bare soil fraction estimate with  $15.0\% \pm 3.9$ . These result would suggest that the information contain in the biomass map is not enough to disantrangle grass to baresoilestimates (Fig. 3B).

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

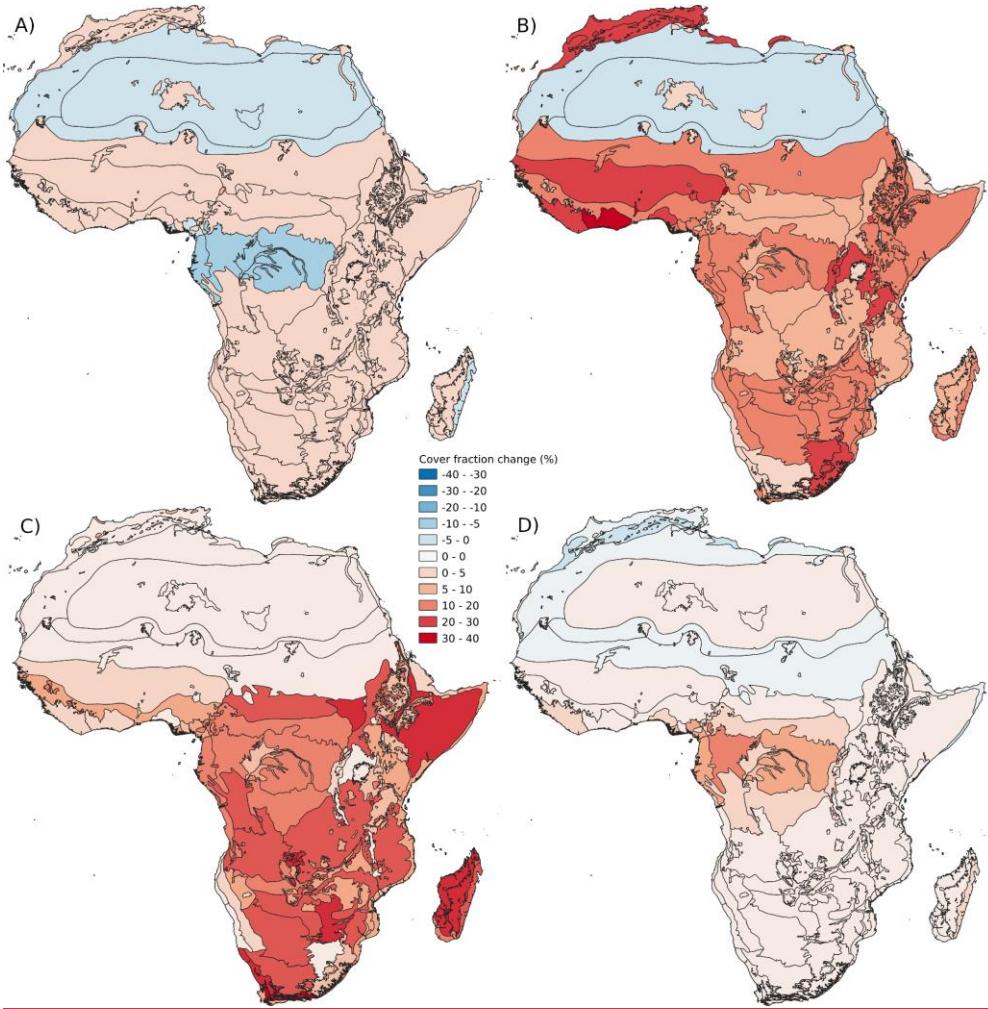
a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

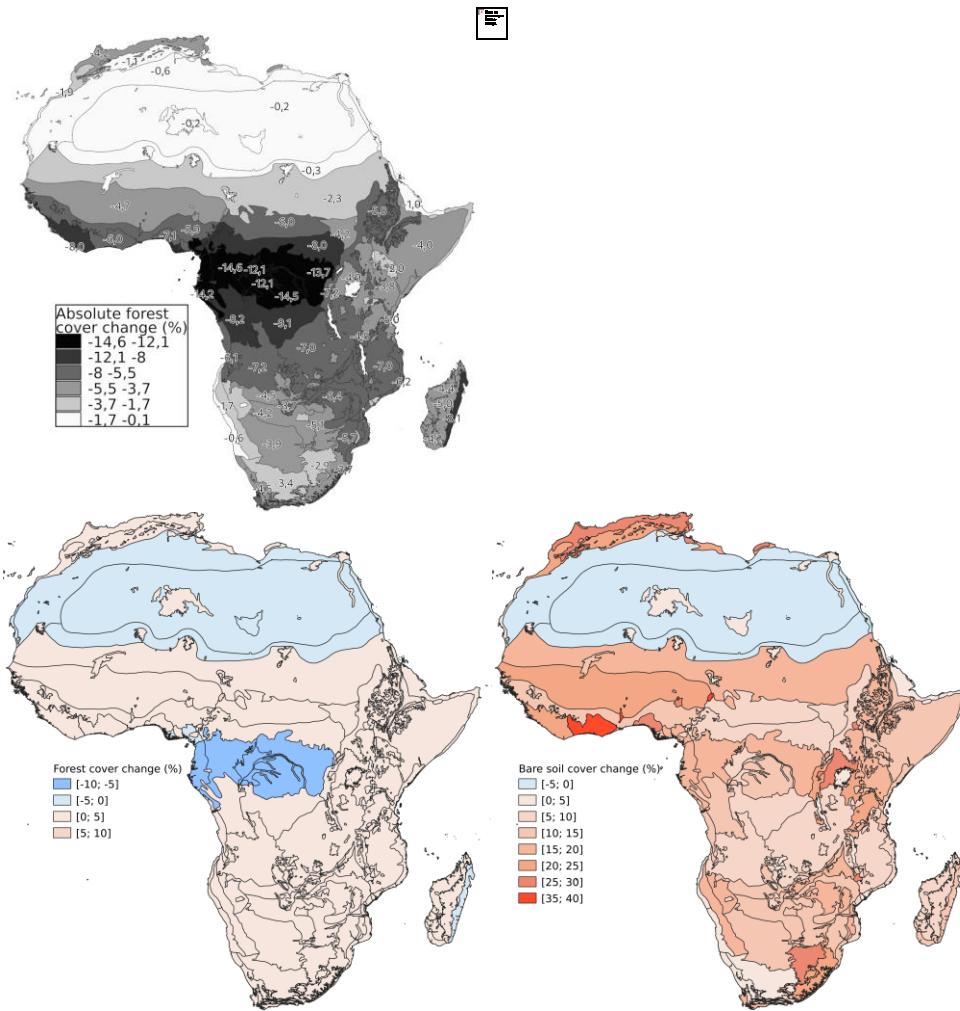


**Figure 3:** Uncertainty in CWT constrained by an AGB map. Absolute change in forest (A) and bare soil (B) cover fraction, where the 2,5% CWT PFT map display more baresoil fraction and less grass fraction

**a mis en forme :** Police :9 pt, Gras, Italique, Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite



**Figure 3:** A map of African's ecoregions showing the absolute change in forest (left) and the bare soil (right) cover fraction percentage (%) between the 2,5 and 97,5 percentile PFT maps. It can be seen as the uncertainty of the newly developed method. High values represent a large uncertainty in the estimation of the true cover fraction. The definition of the cover fraction. C and D represent disagreement estimated as the difference between the CWT based on expert knowledge and the CWT constrained by an AGB map. Disagreement in forest (C) and bare soil (D) is expressed as absolute change (%). High values represent a strong disagreement between the two methods. Black lines delimit the different ecoregion has been taken from according to Olson et al., 2001.

**a mis en forme :** Normal, Espace Après : 10 pt, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

**a mis en forme :** Police :9 pt, Gras, Italique, Couleur de police : Noir

**a mis en forme :** Police :9 pt, Gras, Italique

**a mis en forme :** Police :9 pt, Gras, Italique, Couleur de police : Noir

**a mis en forme :** Police :9 pt, Gras, Italique

**a mis en forme :** Police :9 pt, Gras, Italique, Couleur de police : Noir

**a mis en forme :** Police :9 pt, Gras, Italique

**a mis en forme :** Police :9 pt, Gras, Italique, Couleur de police : Noir

**a mis en forme :** Police :9 pt, Gras, Italique

**a mis en forme :** Police :8 pt, Soulignement , Couleur de police : Noir

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

**Table 2a: Short description, surface2: Surface area (%), share in the continental biomass (%), prior parameters, and posterior median and confidence credible interval values for each of the 1615 land cover types considered in this study. The numbering, description and surface area of each land cover type is based on the ESA-CCI product [Defourny, P., 2019], where its share in the continental biomass is based on a compilation of Bouvet et al 2018.  $\theta_{lc,i}$ ,  $\mu_{lc,i}$  and  $\sigma_{lc,i}$  represent the parameters describing the prior distributions of  $F_{lc,i}$  and  $Bref_{lc,i}$ . Estimation of these parameters is detailed in section 2.3. For each land cover type and each parameter, the 2,5, the mode constrained and the 97,5 percentiles are computed. The We use the mode is a good for the constrained CWT as an approximation of the posterior  $\theta_{lc,i}$ , since the posterior distributions of  $F_{lc,i}$  may be asymmetric**

a mis en forme : Police :9 pt, Gras, Italique, Non souligné
a mis en forme : Normal, Espace Après : 10 pt, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique, Couleur de police : Automatique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Police :9 pt, Gras, Italique
a mis en forme : Couleur de police : Noir
a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

id	UN-LCCS short description	Informations			Priors			Posterior fractions used for the CWTs											
		Surface area (%)	Biomass (%)	$F_{k,w}$	$F_{k,h}$	$\theta_{k,w}$	$\theta_{k,h}$	$\mu_{k,w} \pm \sigma_{k,w}$	$Bref_{k,w}$	$Bref_{k,h}$	CWT 2,5 %			CWT 97,5 %			CWT 50 %		
											$F_{k,w}$	$F_{k,h}$	$\mu_{k,h} \pm \sigma_{k,h}$	$F_{k,w}$	$F_{k,h}$	$F_{k,w}$	$F_{k,h}$		
10	Cropland rainfed	7,6	5	0,01	0,98	0,01		83±3,1		9±0,3	0,19	0,34	0,47	0,13	0,86	0,01	0,14	0,83	0,03
11	Cropland rainfed - Herbaceous cover	3,2	3,3	0,01	0,98	0,01		84±3,1		6±0,2	0,13	0,49	0,38	0,09	0,89	0,02	0,11	0,85	0,04
30	Mosaic cropland (>50%) / natural vegetation (tree/shrub/herbaceous cover) (<50%)	2,3	3,1	0,25	0,74	0,01		85±3,2		4±0,2	0,2	0,77	0,03	0,17	0,83	0,00	0,18	0,81	0,01
40	Mosaic natural vegetation (tree/shrub/herbaceous cover) (>50%) / cropland (<50%)	2,2	1,9	0,5	0,49	0,01		84±3,1		9±0,3	0,26	0,69	0,05	0,22	0,78	0,00	0,24	0,75	0,01
50	Tree cover broadleaved evergreen closed to open (>15%)	6,7	45,1	0,99	0	0,01		416±15,6		9±0,3	0,71	0,02	0,27	0,79	0,1	0,11	0,76	0,01	0,23
60	Tree cover broadleaved deciduous closed to open (>15%)	4,2	8,7	0,7	0,29	0,01		245±9,2		9±0,3	0,27	0,66	0,07	0,23	0,76	0,01	0,25	0,73	0,02
61	Tree cover broadleaved deciduous closed (>40%)	0,4	1,8	0,85	0,14	0,01		252±9,5		9±0,3	0,54	0,44	0,02	0,61	0,3	0,09	0,58	0,4	0,02
62	Tree cover broadleaved deciduous open (15-40%)	10,6	13,1	0,55	0,44	0,01		111±4,2		9±0,3	0,35	0,61	0,04	0,3	0,69	0,01	0,32	0,66	0,02
100	Mosaic tree and shrub (>50%) / herbaceous cover (<50%)	1,8	1,5	0,6	0,39	0,01		85±3,2		9±0,3	0,16	0,77	0,07	0,13	0,86	0,01	0,15	0,84	0,01
110	Mosaic herbaceous cover (>50%) / tree and shrub (<50%)	1,6	1,2	0,4	0,59	0,01		75±2,8		5±0,2	0,07	0,85	0,08	0,05	0,94	0,01	0,06	0,93	0,01
120	Shrubland	13,3	7,7	0,6	0,39	0,01		85±3,2		9±0,3	0,16	0,74	0,1	0,12	0,87	0,01	0,14	0,85	0,01
130	Grassland	6,5	1,5	0,01	0,98	0,01		42±1,6		3±0,1	0,11	0,56	0,33	0,07	0,92	0,01	0,09	0,9	0,01
150	Sparse vegetation (tree/shrub/herbaceous cover) (<15%)	1,6	0,2	0,1	0,2	0,7		12±0,5		3±0,1	0,04	0,45	0,51	0,15	0,01	0,84	0,1	0,19	0,71
153	Sparse herbaceous cover (<15%)	1,1	0,1	0,01	0,29	0,7		22±0,8		3±0,1	0,1	0,29	0,61	0,01	0,97	0,02	0,02	0,81	0,17
160	Tree cover flooded fresh or brackish water	0,7	3,5	0,75	0,24	0,01		386±14,5		9±0,3	0,6	0,39	0,01	0,69	0,27	0,04	0,65	0,34	0,01

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

id	UN-LCCS short description	Informations			Priors			Posterior fractions used for the CWTs										
		Surface area (%)	Biomass (%)	$F_{lc,w}$	$F_{lc,b}$	$F_{lc,h}$	$Bref_{lc,w}$	$Bref_{lc,h}$	CWT 2.5 %			CWT 97.5 %			Refined CWT			
				$\theta_{lc,w}$	$\theta_{lc,b}$	$\theta_{lc,h}$	$\mu_{lc,w} \pm \sigma_{lc,w}$	$\mu_{lc,h} \pm \sigma_{lc,h}$	$F_{lc,w}$	$F_{lc,b}$	$F_{lc,h}$	$F_{lc,w}$	$F_{lc,b}$	$F_{lc,h}$	$F_{lc,w}$	$F_{lc,b}$	$F_{lc,h}$	
10	Cropland rainfed	7,6	5	0,01	0,98	0,01	83±3,1	9±0,3	0,19	0,34	0,47	0,13	0,86	0,01	0,14	0,83	0,03	
11	Cropland rainfed - Herbaceous cover	3,2	3,3	0,01	0,98	0,01	84±3,1	6±0,2	0,13	0,49	0,38	0,09	0,89	0,02	0,11	0,85	0,04	
30	Mosaic cropland (>50%) / natural vegetation (tree/shrub/herbaceous cover) (<50%)	2,3	3,1	0,25	0,74	0,01	85±3,2	4±0,2	0,2	0,77	0,03	0,17	0,83	0,00	0,18	0,81	0,01	
40	Mosaic natural vegetation (tree/shrub/herbaceous cover) (>50%) / cropland (<50%)	2,2	1,9	0,5	0,49	0,01	84±3,1	9±0,3	0,26	0,69	0,05	0,22	0,78	0,00	0,24	0,75	0,01	
50	Tree cover broadleaved evergreen closed to open (>15%)	6,7	45,1	0,99	0	0,01	416±15,6	9±0,3	0,71	0,02	0,27	0,79	0,1	0,11	0,76	0,01	0,23	
60	Tree cover broadleaved deciduous closed to open (>15%)	4,2	8,7	0,7	0,29	0,01	245±9,2	9±0,3	0,27	0,66	0,07	0,23	0,76	0,01	0,25	0,73	0,02	
61	Tree cover broadleaved deciduous closed (>40%)	0,4	1,8	0,85	0,14	0,01	252±9,5	9±0,3	0,54	0,44	0,02	0,61	0,3	0,09	0,58	0,4	0,02	
62	Tree cover broadleaved deciduous open (15-40%)	10,6	13,1	0,55	0,44	0,01	111±4,2	9±0,3	0,35	0,61	0,04	0,3	0,69	0,01	0,32	0,66	0,02	
100	Mosaic tree and shrub (>50%) / herbaceous cover (<50%)	1,8	1,5	0,6	0,39	0,01	85±3,2	9±0,3	0,16	0,77	0,07	0,13	0,86	0,01	0,15	0,84	0,01	
110	Mosaic herbaceous cover (>50%) / tree and shrub (<50%)	1,6	1,2	0,4	0,59	0,01	75±2,8	5±0,2	0,07	0,85	0,08	0,05	0,94	0,01	0,06	0,93	0,01	
120	Shrubland	13,3	7,7	0,6	0,39	0,01	85±3,2	9±0,3	0,16	0,74	0,1	0,12	0,87	0,01	0,14	0,85	0,01	
130	Grassland	6,5	1,5	0,01	0,98	0,01	42±1,6	3±0,1	0,11	0,56	0,33	0,07	0,92	0,01	0,09	0,9	0,01	
150	Sparse vegetation (tree/shrub/herbaceous cover) (<15%)	1,6	0,2	0,1	0,2	0,7	12±0,5	3±0,1	0,04	0,45	0,51	0,15	0,01	0,84	0,1	0,19	0,71	
153	Sparse herbaceous cover (<15%)	1,1	0,1	0,01	0,29	0,7	22±0,8	3±0,1	0,1	0,29	0,61	0,01	0,97	0,02	0,02	0,81	0,17	
160	Tree cover flooded fresh or brackish water	0,7	3,5	0,75	0,24	0,01	386±14,5	9±0,3	0,6	0,39	0,01	0,69	0,27	0,04	0,65	0,34	0,01	

### 3.2.1 Comparison between the refined and the original PFT maps

Nonetheless, in a classic simulation experiment where ORCHIDEE is invoked, the most likely common values of  $F_{lc,w}$ ,  $F_{lc,b}$ ,  $F_{lc,h}$  will be used. These likely The most common values of  $F_{lc,w}$ ,  $F_{lc,b}$ ,  $F_{lc,h}$  are represented given by the mode of the posterior distribution (referred “constrained CWT” in this study as “refined CWT”). This analysis cover Table 2.a). The mode was used to show the disagreement difference between the original and the refined constrained PFT maps. (Fig. 3C-D). The mean change difference in forest cover fraction between the two maps are -11,7% (-14,8%). Large prior (original) and the constrained PFT maps is -15±12% (Fig. 3C). Largest disagreement between the refined and original maps was observed over the Somali Acacia-Commiphora Bushlands and Thickets and the Kalahari Xeric Savanna where forest cover fraction was found to be on average 32±1% lower in average for the refined constrained PFT maps (Fig. 4a-b3C). The guinea forest showed a 6,5 to 12,5%

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

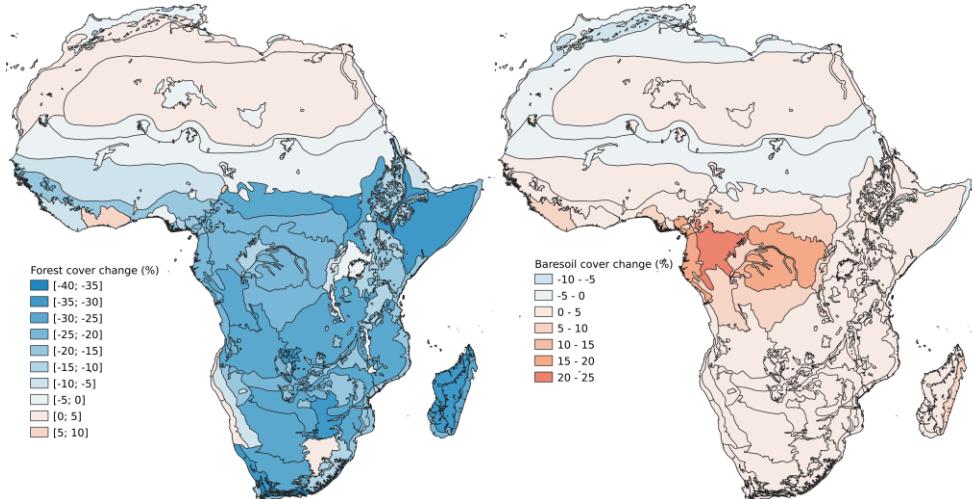
a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

higher forest**Bare soil** cover fraction for the refined changes on average by  $3.1 \pm 0.5\%$  (Fig. 3D). The constrained PFT maps map has on average  $16 \pm 4\%$  more bare soil cover fraction over the Congo Basin than the original map (Fig. 3C).



**Figure 4:** Forest (left) and baresoil (right) cover fraction change from the original PFT map. PFT map generation is described in the section 2.4.1.

### 3.3 EffectUncertainty propagation of the PFT maps on the aboveground biomass and visible albedo estimates from ORCHIDEE simulations

PFT maps are essential boundary conditions of land surface models because they condition the spatial distribution of various ecosystem states-properties (i.e., carbon content, albedo, water-carbon-energy fluxes, etc). When tested with the ORCHIDEE tags 2.0 (rev 6592), the absolute difference in biomass stock between the 2,5 and 97,5 percentile maps was  $4,80,5 \pm 5,7$  t/ha<sup>+/-</sup> (not shown). The small differences in biomass between the original simulations (Fig. S2) can be explained by a modest cover fraction change from -12,2% to 1,3% in tropical rainforest (UN LCCS 50 and 160), respectively for the 2,5% and the 97,5% PFT maps compared to the original PFT map (Fig. 4). When the PFT maps are propagated through ORCHIDEE, the aboveground biomass changes for these ecoregions result in -174A representing 0,2 t/ha<sup>+/-</sup> and 3,9 t/ha<sup>+/-</sup>, respectively from the 2,5% and the 97,5% PFT maps compared to the original PFT map (Fig. 5a-b). The difference in AGB estimates between the original and refined PFT maps was the largest over Madagascar (-52,3 t/ha<sup>+/-</sup> in 2,5%), the Eastern Guinean Forests (+13,5 t/ha<sup>+/-</sup> in 97,5%), and the Northwestern Congolian Lowland Forests (-33,9 t/ha<sup>+/-</sup> in 2,5%) (Fig. 5a-b).

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

At the ecoregion scale, the largest difference between the % of cover fraction (Fig 4C). Notable exception is the Congo basin where different PFT maps could result in AGB estimates that differ by 18 t/ha (Fig. 4A) for a 6.5% difference in the forest cover (Fig 3A). Different PFT maps make the average visible albedo range from  $0.081 \pm 0.055$  to  $0.083 \pm 0.055$ . The largest uncertainty for the visible albedo simulated with ORCHIDEE tags 2.0 (rev 6592) rev2.1 and initialized with the 2,5 and 97,5 percentile of the distribution of PFT maps and the simulated albedo for the original PFT map was found in Madagascar and ranged from 0,014 and 0,010 respectively from the 2,5% and the 97,5% over the Nigerian lowland forest (0,158) and West Sudanian Savanna (0,107) (Fig. 5e-d). Nonetheless, the overall effect of the different PFT maps on the albedo is less than 0,0005,4B which represent a 24% to 11% change in forest cover respectively. The sensitivity is the highest in the western Congo basin with 1,4% of albedo/% of cover fraction. In contrast, West Sudanian Savanna possesses a low sensitivity with 0,5%. To summarise, we found that a smaller forest to bare soil transition uncertainty can drastically change the albedo of an ecoregion than a larger uncertainty in the grassland/cropland to bare soil transition.

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

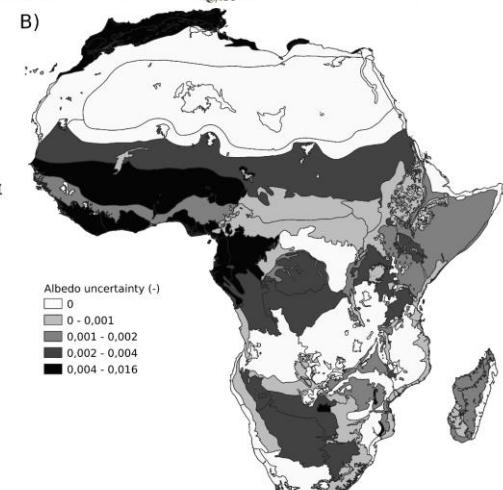
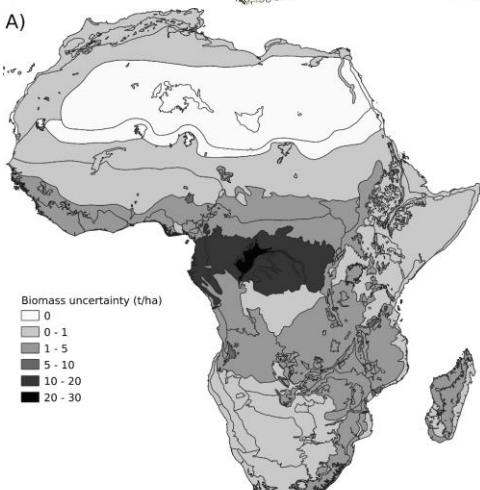
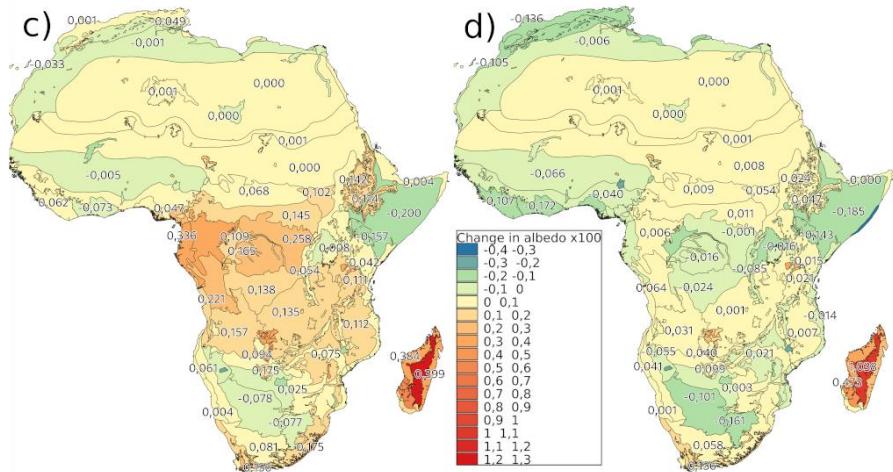
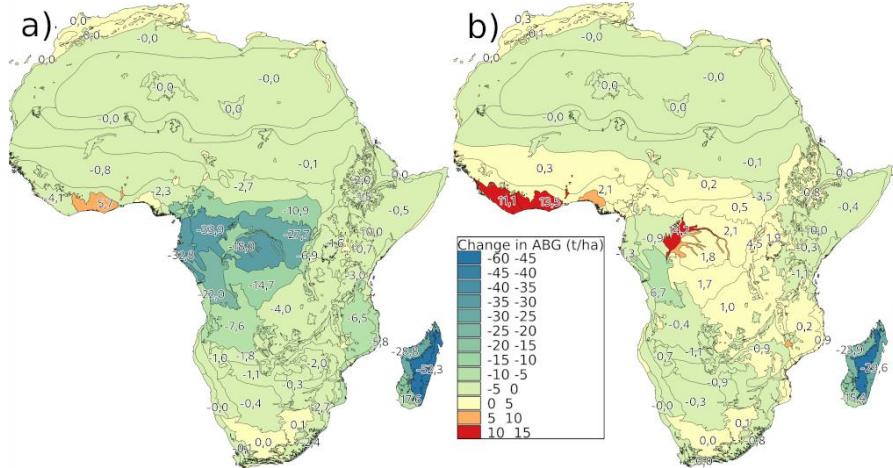
a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Texte 1

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite



**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

**a mis en forme :** Couleur de police : Noir

**a mis en forme :** Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

**Figure 54:** Confident interval propagation of the PFTs maps into AGB and visible albedo simulated by ORCHIDEE. The left panels represent the difference in (A) uncertainty propagation into AGB, or and (B) uncertainty propagation into visible albedo from the difference between the 2,5% CI and the 97,5% PFT map and the original PFT map. The right panels represent the difference undefined by the 97,5% CI PFT map and the original PFT map, optimisation procedure, uncertainty propagation index (eq. 16) for AGB (C) and b) are above ground biomass change in t.ha<sup>-2</sup>, and c) and d) are change in visible Albedo \* 100. visible albedo (D).

a mis en forme

## 4 Discussion

### 4.1 Discretizing vegetation

Irrespective of the data products, the methods, and the model used, discretizing vegetation comes with its own challenges. Representing transitions of ecosystems by discretizing the vegetation into land cover type classes (Sankaran et al., 2005), for example, can lead to systematic errorsuncertainties since all pixels that belong to the same land cover class will get the same vegetation cover fractions (see 4.1.3) in the cross-walking table. (see 4.1.3). This approach articulates a key assumption underlying the PFT-approach, i.e., that only one life form survives and thus dominates the vegetation due to competition for nutrients, light and water (Hutchinson et al., 1961). However, the Savanna ecosystem, for example, is characterised by the coexistence of trees, shrubs and grasses which has been explained by interactions between vegetation, rainfall, fire, and browsing regimes (Egentler and Sherratt, 2020). This makes savannas one of the most difficult ecosystems to classify in a land cover type and subsequently convert it into a PFT map.

Over Africa, land cover classes such as shrubland (UN-LCCS 120) represent a wide range of ecosystems, from sparse xeric shrubland composed of small bushes, e.g., *Penzia incana* (Thunb.) Kuntze, grasses, e.g., *Sip Agrostis agrostis* spp. such as found in Karoo desert, to dense thicket composed by succulent, e.g., *Portulacaria afra* Jacq. and spinescent shrubs (~3m tall) (Mills, et al., 2005). Combining land cover types and biomass maps showed that the shrubland pixels in Africa more often resemble sparse xeric shrubland than dense thickets. Improving the ability to simulate land surface properties of shrublands in a changing world, especially in Africa where shrub encroachment is an important land cover dynamic (Wigley et al., 2010, Buitenhof et al., 2012, O'Connor et al., 2014), is likely to benefit from a more detailed representation of shrublands in land surface models. A first step could be to represent shrubs as small trees, as was tested with the ORCHIDEE model for arctic ecosystems (Druel et al., 2017), but ultimately shrub density, largely controlled by the control of precipitation (on plant density (Rietkerk et al., 2002)) should also be modeledmodelled.

Another major challenge with discretizing vegetation is how degraded ecosystems should be classified. From a modelling point of view, they should be classified as the land cover type that occurred prior to the degradation and the cause of the degradation. e.g., fire, grazing, erosion, should be explicitly accounted for in the land surface model. This ideal strongly differs from the current approach in which the degraded vegetation is classified as if the degraded vegetation is in its natural state.

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme

a mis en forme

a mis en forme

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

Even when having the correct PFTs, the current approach ~~will~~would fail to simulate the observed biomass if degradation occurred. As an alternative, the PFT map could duplicate all PFTs to distinguish between a PFT in its natural state and in its degraded state. This approach in which degradation is accounted for in the PFT maps would, however, reduce degradation to a binary problem rather than addressing its continuous nature.

#### 4.2 Knowledge gain from using the AGB map

In the absence of an AGB map, previous efforts to build cross-walking tables [\(Poulter et al., 2015\)](#) had to rely in part on expert knowledge. That generation of cross-walking tables can be considered as the best-available-knowledge in the absence of AGB data or other information on the land surface cover. The method developed and demonstrated in this study mostly relies on data but comes with its own assumptions and statistical complexities. The key assumptions are that: (1) previous cross-walking tables [\(Poulter et al., 2015\)](#) are a reliable source to set the prior distribution for PFT cover, (2) the biomass map [\(Bouvet et al., 2018\)](#) is a reliable source to set the prior distribution of the reference biomasses, and (3) the land cover classification contains homogeneous land cover types [\(Defourny, P. et al., 2019\)](#). A key question is thus whether the added complexity justifies the knowledge gained by jointly assimilating a land cover and a biomass map when producing a CWT? Ideally this question should be addressed by assessing the reduction of the ~~confident~~credible interval associated to the posterior distribution of the PFT map when using the AGB map to constrain the CWT (in comparison to a prior when no AGB is used). However, ~~the present generation of CWT without AGB information, does not come with a distribution (except the attempt in Hartley et al., (2017))~~, calling for an alternative approach to assess the knowledge gain. Given that the prior distribution of the cover fraction was based on the previous CWT, the difference between the prior and the posterior distributions can be considered as the knowledge gained from using AGB information. Following this reason, the question we seek to answer is: "Is the cover fraction used by the original cross walking table falling outside the 95% ~~confidenee~~credible interval of our posterior estimate?"

If the answer is no, the biomass map is more likely in agreement with the previous effort to estimate the original cross walking table. If the answer is yes, adding the information contained in the satellite-based biomass maps is most likely in strong disagreement with the previous effort to estimate the original cross walking table. The original CWT has a global extent, and the ~~refined~~constrained CWT is only valid for Africa. Therefore, knowledge gains should be carefully interpreted as they may reflect trade-offs that had to be made previously to construct a global rather than a regional CWT. Knowledge gains were assessed for: "croplands", "dense evergreen forests", "woodlands and savannas", and "xeric shrublands and grasslands" separately.

##### 4.2.1 Croplands (UN-LCCS 10, 11, 30, 40).

Despite the cover fraction of woody vegetation on croplands being close to none in the original CWT, this study found that the four land cover types associated with croplands, UN-LCCS 10, 11, 30, 40 are in fact covered ~~by~~with 11% to 24% woody

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

vegetation ranging from 15% to 30% (range of median; Table 2). This large difference in the presence of woody vegetation on croplands is also reflected in the biomass data, which suggest two distinct but co-existing agricultural systems in Africa, i.e., one system with a low biomass and one around with a higher biomass.

The agricultural system with the low biomasses likely represents annually replanted crops such as millet, sorghum, wheat, sweet potatoes or cassava (FAO), with a maximum reported biomass between 10 and 15 t/ha<sup>+</sup> for high-input cropping associated with commercial production of cassava and sweet potatoes. These values are in line with values estimated as reference biomass (see 2.3.2). Nonetheless, 97% of total cropland area Africa is rainfed (Calzadilla et al., 2009) and most of Africa's agricultural land is used for subsistence or small-scale farming associated with low-input cropping which explains why the actual average biomass estimate from the CESBIO map for cropland is between 1,98±2,0,65±0,7 t/ha<sup>+</sup> (Fig. 2) and thus considerably lower than the potential production.

The high biomass agricultural system which is estimated at 58,6±10,283±3 t/ha<sup>+</sup> in the CESBIO map (Fig. 2) likely includes plantations for coffee, rubber, fruits as well as shelter trees and forest remnants (FAO). Permanent croplands do not have their own land cover type in the UN-LCCS or in ORCHIDEE. In the absence of a dedicated PFT, these agricultural systems could be better simulated with a woodland fraction ranging from 13% to 34% (see table 2), than is currently done with the standard CWT. The mixture of bare soil, herbaceous vegetation and woody vegetation, makes it challenging to discretize African croplands into the current PFTs (Table 1). Moreover, small changes in the woody reference biomass for high biomass agricultural systems lead to large changes in cover fractions of herbaceous vegetation and bare soil ratio. Without constraint reference biomass estimates, total biomass alone does not sufficiently constrain the share of woody vegetation. For the time being, high biomass agricultural systems could be with a woodland fraction ranging from 9,0% to 26% (Table 2). Although this could be an acceptable solution for biomass and albedo simulations, it will underestimate the agricultural production in the region.

#### 4.2.2 Tropical rainforest (UN-LCCS 50, 160).

The woody cover fraction of tropical rainforest in the original CWT is close to 90%. The original woody cover % and falls inside/outside the confidence/credible interval of the posterior estimates, possibly because the 95% confidence interval is considerable and ranges from 80 i.e., 71 to 100%. Strictly speaking, using a biomass map in addition to a land cover map did not result in knowledge gain concerning the 79%. This lower cover fraction of tropical rainforest. Nevertheless, the considerable range in cover fraction indicates that from many of the pixels classified as tropical rainforest that do not all achieve/reach the reference biomass of 371±14416±16 t/ha<sup>+</sup> (Fig. 2). The reference derived from the biomass map matches the AGB observed at field plots of intact forests in the Congo basin (Lewis et al., 2013), but the large variation value in bare soil cover fraction for these land cover types may thus reflect wide-spread degradation of the forests in the region (Tyukavina et al., 2018) or an unrepresentative too high reference biomass (Kearsley et al., 2013).

#### 4.2.3 Tropical moist deciduous forest/woodland/and savanna (UN-LCCS 61, 60 and 62).

The woody cover fraction of the tropical moist deciduous forest ranged between 55.45% and 85.75% in the original CWT.

Refining the CWT by the use of using AGB information narrowed shifts this range to between 55.27% and 68.58%. For savanna (UN-LCCS 62) and woodland (UN-LCCS 60), the original cover fractions are within the refined constrained 95% CI. For woody cover, the fraction of moist deciduous forest (UN-LCCS 61) decreased from 85% to 65%.58%. We observe an overall decrease for the woody cover fraction since the reference biomass is much higher than the actual biomass of most of the pixels.

Although the reference biomasses used in this study are in line with previously reported values (Carreira et al., 2013), there are two ways to interpret the continuously decreasing biomass when moving from a forest, over a woodland towards savanna. The original CWT considered these three land cover types as a single PFT. Differences between potential the original and actual biomass are the outcome of land use and are reflected in largely different woody cover fractions between these land cover types. the constrained CWT is considerable. The original CWT starts from the view that all ecosystems (except croplands) are in their natural state. The AGB map, however, does not contain any evidence in support of this view and but rather suggests suggests that each 50% of these land cover types comes with an own the savanna (UN-LCCS 62) are 65% below their reference biomass, i.e., 197±30 t.ha<sup>-1</sup>. Likewise, 50% for dry forest, 62±14 t.ha<sup>-1</sup> for woodland and 22±12 t.ha<sup>-1</sup> for savanna (Fig. 2). Such a gradient in woodland (UN-LCCS 60) are 71% below their reference biomass could be justified by a climatological gradient. (Fig. 2). The large range in reference biomasses compensated the range in actual biomass resulting in a relatively small range in woody cover fractions. Given the considerable AGB map thus suggests wide-spread degradation of these land-cover types [ecosystems which are in a highly anthropized region (Mitchard et al., 2013), the reality is likely a combination]. Uncertainty coming from the reference biomasses could be reduced by field observations at the ecoregion or finer spatial scales.

For deciduous forest, however, the difference in cover fraction of degradation superimposed on a climate gradient woody vegetation between the original CWT and the constrained CWT could also be explained by an inaccurate estimation of the reference biomass due to a too coarse definition of the deciduous woody vegetation ranging from deciduous forest, over woodlands to savannas which are composed by different dominant tree species, with different biomasses (Sawadogo et al., 2010).

#### 4.2.4 Xeric shrubland (UN-LCCS 100, 110, 120).

The woody cover fraction of xeric shrublands and grasslands ranged between 40 and 60% in the original CWT. Accounting for the information contained in the AGB map significantly decreased the woody cover fraction range toward 85.0 and 49.16%.

Indeed, shrubs which represent a large part of the xeric shrublands were originally classified as woody vegetation for the ORCHIDEE model (i.e., when moving from the generic PFTs to the ORCHIDEE-specific PFTs; see section 2 and

**ORCHIDAS**). This assumption is true from an ecological point of view but in a simplified world like in land surface models, xeric shrubland have has an aboveground biomass that resembles that of cropland and grassland (Fig. 2). By overlaying the land cover type and aboveground biomass maps, 37% of the African shrublands were found to be degraded with a biomass of  $2.7 \pm 1.5 \text{ t/ha}^+$ , 54% were found to be intact with a biomass of  $22 \pm 19 \text{ t/ha}^+$  and 9% of the shrublands are thickets with a biomass of  $68 \pm 11 \text{ t/ha}^+$ . This is in line with other aboveground biomass estimates from remote sensing products (Saatchi et al., 2011; Mitchard et al., 2013; Avitabile et al., 2016) and in situ measurements where shrublands, degraded thicket, and intact thicket in south Africa accumulated 3, 24 and 102 t·ha<sup>+</sup> of biomass respectively (Mills, et al., 2005). These findings suggest that in the model world, xeric shrubland is best represented by a large fraction of herbaceous plant functional groups, when the overall objective is to model AGB.

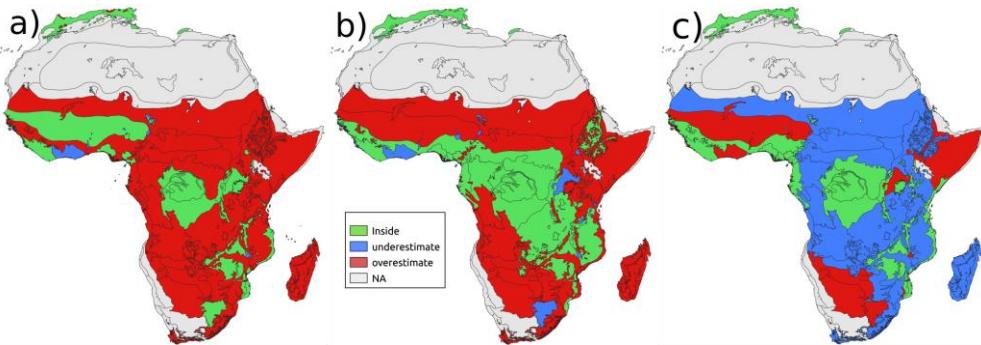
#### 4.2.4 Sparse vegetation (UN-LCCS 150, 153).

The constrained cover fraction estimates are in line with the original CWT for UN-LCCS 150 which represent the most common class of sparse vegetation. The constrained cover fraction for UN-LCCS 153 has a larger herbaceous i.e., 29 to 97%, then the bare soil cover fraction, i.e., 2,0 to 61% contrary to the original CWT. The herbaceous cover fraction could be overestimated if a too low reference biomass was used. A reference biomass of 3,0 t/ha was used and is acceptable compared to the reported biomass for the Succulent and Nama Karoo Biomes ranging from 0.5 to 7.6 t/ha (Rutherford, 1978; Rutherford and Westfall, 1986). Given the current lack of reference biomass observations, disagreement between the original and constrained CWT could be resolved by using an independent estimate of bare soil fraction.

### 4.3 Consequences for land surface modelling

#### 4.3.21 Which land cover types affect the biomass estimate?

In Tropical Rainforest (UN-LCCS 50, 160) and Deciduous Moist Forest (UN-LCCS 60, 61 and 62), the AGB obtained by driving ORCHIDEE with the original PFT map falls inside the confident interval simulated by driving ORCHIDEE with the refined PFT maps (Fig. 6). This is clearly not the case in the rest of the African ecoregions where the original PFT map systematically overestimates the 95% CI of simulated AGB obtained by using the refined PFTs maps. In general, the overestimation with the original PFT map is explained by an overall reduction in the forest-cover fraction in the refined PFT maps.



**Figure 6:** Is the original PFT map fall inside the 95% confidence interval of the refined PFT maps (a) when used estimates (30 to drive ORCHIDEE to simulate AGB (b) and visible Albedo (c). For ecoregions colored 40% resulted in small disagreement in green, the AGB simulated making use of the original PFT map falls within the simulated 95% AGB range obtained by driving ORCHIDEE with the refined 2,5% and 97,5% PFT maps. Blue shows biomass, i.e., <1.0 t/ha in regions where the original PFT maps results in a significantly lower than the refined PFT maps. Red shows regions where the original PFT maps results in a significantly higher than the refined PFT maps. Grey indicates regions where the simulated AGB is less than 1 t.ha<sup>-1</sup>.

Compared to the original maps the refined maps prescribe a 28,4% to 32,4% lower tree cover for shrubland ecoregions with little precipitation like Somali Acacia-Commiphora Bushlands and Thickets and the Kalahari Xeric Savanna. Propagating these changes in cover fraction into simulated AGB, resulted in only small changes in carbon stocks (<1 t.ha<sup>-1</sup>). This counter-intuitive result is explained by the growth processes simulated in ORCHIDEE. Under xeric climate conditions ORCHIDEE simulates low tree biomasses (<2,0 t.ha<sup>-1</sup>) because the low precipitation and subsequent plant water availability results in a continuous high tree mortality. In the refined maps, about 1/3 of the trees are replaced by grasslands which given the plant available water survive and even grow up to biomass of ~1,5 t.ha<sup>-1</sup>.

Humid Mixed Cropland/Forest ecoregions like the eastern Guinean forests were systematically underestimated by the original PFT map. Indeed, compared to the original PFT map, the refined PFT maps prescribe a higher forest cover fraction for the cropland land cover type in order to include permanent tree crops like cacao, coffee and rubber plantation. Increasing the forest cover in humid regions results in a strong increase in the simulated AGB because in ORCHIDEE forests under these climates reach high (> 100 t.ha<sup>-1</sup>) biomasses. To conclude, underestimating nonetheless, in forest ecoregions like the eastern Guinean forests or in the Congo basin, where the sensitivity to a change in the cover fractions ranged from 1,0 to 5,0 t/ha/% and had a considerable impact on the simulation since a 15% uncertainty in the bare soil fraction may lead to a 75 t/ha uncertainty of the biomass in the tropical forest of the Congo basin. Underestimating the forest cover in humid ecoregions will have a much larger consequence on the simulated AGB than overestimating the forest cover in xeric ecoregions. The uncertainty surrounding the land cover fractions should thus be further reduced for the land cover types that already come with the lowest uncertainty, i.e., the forests.

a mis en forme : Police :Non Italique

a mis en forme : Police :Non Italique

a mis en forme : Police :Non Italique

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3,13", Centré + 6,27", Droite

#### 4.3.32 Which land cover types affect the albedo estimate?

Overestimating the biomass is likely to come with overestimating the leaf area which in turn will result in underestimating the albedo because the reflectivity of leaves is often lower than the reflectivity of the soils it covers [Oke 2002]. Hence, Fig. 6e is expected to mirror Fig. 6b. This is not the case for the Somali Acacia Commiphora Bushlands and Thickets and Sahelian ecoregions were replacing the forest by better growing grassland, both the biomass and albedo are decreasing. As for AGB, uncertainties in land cover fractions are only partly reflected in the uncertainties of the visible albedo. Dampening is caused by the fact that the reflectivity of grassland (0.06), cropland (0.06) are close to the leaf reflectivity of a forest (0.03 to 0.04) compared to bare soils reflectivity (0.1 to 0.25 depending on the colour of the soil) in ORCHIDEE. By increasing the bare soil cover fraction, the albedo will increase accordingly but changing forest into grassland will not drastically change albedo. The most sensitive area is the western tropical forest in the Congo basin for which a 15% change in bare soil cover fraction may trigger a 15% change in the visible albedo (fig 3C). Similar as for AGB, the uncertainty surrounding the land cover fractions of the forested land cover types should be further reduced to reduce the uncertainty of the model simulations.

#### 4.4 Outlook

In this study a single biomass map was used as this enabled keeping the focus on the method itself. Nevertheless, other biomass products are available (Saatchi et al., 2011; Baccini et al., 2012; Avitabile et al., 2016; Santoro et al., 2020; 2021) and could have been used. Repeating this study for each of these biomass products would add another source of uncertainty to the cross-walking table. Owing to the method presented in this study, this uncertainty could then be propagated into the PFT map and all the way up to the simulated biomass, albedo -as done in this study for one biomass product- and other land surface properties. Considering different biomass products would give an insight of the impact of satellite-based biomass estimates on the discretisation of the vegetation and by extension surface properties as estimated by land surface models. Likewise, a single land cover map has been used in our analysis, but other products are available as well (Copernicus, UN-spider, Li et al., 2020). By using different land cover maps, one could quantify the uncertainty in the land cover classification and propagate it to evaluate its impact on the simulated land surface properties.

Compared to other continents, the Africa vegetation has been documented by relatively few quantitative observations (Mills, et al., 2005; Saatchi et al., 2011; Asner et al., 2012; Réjou-Méchain et al., 2015+). Hence, it is the continent where remote sensing data could largely enhance our knowledge on the issue. Recent high-resolution satellite observations bear the promise to significantly reduce the confident credible interval around the aboveground carbon stock to estimate the CO<sub>2</sub> emissions from tropical forests (Hansen et al., 2013; Bouvet et al., 2018; Defourny et al., 2019; Buchhorn et al., 2020) but land surface models will need to be ready to routinely assimilate these data to fully benefit from the information contained in biomass maps. This study demonstrated one way of how satellite-based biomass data can help modellers to refine constraint the initialization process by means of refining the cross-walking tables that are used to map land cover classes derived from

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme

a mis en forme

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

satellite observations into PFT maps. Nevertheless, biomass maps could be used for applications other than model initialization (this study), including model parameterisation and model evaluation.

The biomass map could be used to ~~optimize~~optimise model parameters related to growth, turnover and mortality to better simulate the vegetation biomass for the different PFTs. The evaluation stage could benefit from the biomass maps by benchmarking the model results against observed relationships between biomass-climate and biomass-land-use to better distinguish and simulate the difference between actual and potential biomass (Sankaran et al., 2005). Although the availability of several biomass products makes it possible to use one product to inform the cross-walking tables and another product to evaluate the simulated surface properties, the magnitude of present-day differences between biomass products (Mitchard et al., 2013) is expected to result in major inconsistencies when different biomass products are used for different purposes (e.g., assimilation, parameterization, evaluation) into a single analysis. In this study, less than 0,01% (see 2.3.1) of the information contained in the biomass map was used to ~~refine~~constraint the cross-walking table and none was used to ~~optimize~~optimise model parameters. The simulated biomass (Fig. S2) remains, therefore, largely independent from the biomass map which implies that a single biomass map can be used for land cover optimisation (as in this study), and in a second step for parameter optimization or model evaluation.

With an increase in resolution of the land cover map comes a decrease in the reliance on the cross-walking tables. Cross walking tables will no longer be required once the resolution will be high enough (around 10 x 10 m) such that each pixel contains a single vegetation type equivalent to a single PFT classification used by ~~LSM~~Land surface models (Li et al., 2020). No longer having to rely on cross walking tables would likely reduce the ~~confident interval~~width of the credible intervals of the PFT map. As there would no longer be a need to estimate woody and herbaceous fractions, there would no longer be a need for the information contained in the biomass map. It will then be feasible to solely use biomass maps to better parameterize the processes that contribute to simulating the reference biomass. It should be noted, however, that higher resolutions will not solve the basic challenge of discretizing vegetation. High resolution land cover maps would split structurally complex ecosystems, for example savannas, into a pure forest fraction and a pure grassland fraction. This would ~~overlook~~overlook the interactions between the grasses and the trees which are among the defining ecological characteristics of a savanna.

Finally, we should note that other satellite-derived products than the AGB could be used to ~~refine~~constraint the mapping of the land cover classes into model PFTs (i.e., CWT). For instance, the global tree cover fraction map, at 30-meter resolution, from Hansen et al. (2013) could also be used to ~~refine~~constraint the fraction of ~~tree PFTs~~bare soil within each land cover class (as ~~it like what~~ was done in this ~~paper~~study with the AGB map).

#### 4.5 Conclusion

This study demonstrates how an aboveground biomass map could be used to constrain a cross-walking table that enables remapping land cover types derived from satellite-observations into plant functional types used as a boundary condition in land surface models. Given that previous cross-walking tables did not report uncertainties as they were mostly based on expert-

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

knowledge, it remains unclear how much the use of an additional constraint really improved the cross-walking tables. Nevertheless, the considerable uncertainties remaining in the cross-walking table that made use of the aboveground biomass map suggests that total biomass map should be complemented with a bare soil map to better constrain the cross-walking table. Likewise, the reference biomass for both herbaceous and woody vegetation need to be constrained to at least the ecoregion scale to avoid underestimating or overestimating bare soil fractions. The method developed in this study helped to estimate the uncertainty of cross-walking tables which can now be used to benchmark further methodological developments. Moreover, the method identified bare soil cover fraction would be required to reduce the uncertainty of future cross-walking tables and the plant functional type maps they generate.

## 5 Acknowledgements

This study was primarily financed by the French space agency, Centre National d'Etude Spatiale (CNES), underthrough the specific program "BIOMASS-Valorisation", with " project from the TOSCA research program which contributed to the funding of Guillaume Marie and Cécile Dardel. The Marie Skłodowska Curie Fellowship CLIMPRO (MSCA-Fellowship EU 895455) also contributed partly funded Guillaume Marie.

## 6 Data availability

- CESBIO African AGB map. Biomass map of Africa created by CESBIO can be downloaded on-demand at <https://www.theia-land.fr/en/product/african-biomass-map>. It consists of a GIF file in which Africa is spatially discretized in pixels of 1x1km. The unit is a tonneton of dry mass per hectare (t/ha). Contact person: thuy.leto@alexandre.bouvet@cesbio.cnes.fr
- LandThe land cover map is freely available here from: <http://www.esa-landcover-cci.org>.
- Ecoregion The ecoregion map use follows the work of Olson et al. 2001. This map used in this study, is freely available here from: <https://databasin.org/datasets/68635d7c77f1475f9b6c1d1dbe0a4c4c/>.

## 7 Code availability

- All R scripts and ORCHIDEE tags 2.0 (rev 6592) source code areis available at: <https://zenodo.org/badge/latestdoi/345907299> or DOI: 10.5281/zenodo.4785328 DOI: 10.5281/zenodo.4785328

■ ORCHIDEE tags 2.0 (rev 6592) code is also available at [from](https://forge.ipsl.jussieu.fr/orchidee/wiki/GroupActivities/CodeAvailabilityPublication/ORCHIDEE_tags_2.0_gmd_2021_Africabrowser/tags/ORCHIDEE_2_1/tags/ORCHIDEE_2_1)  
[https://forge.ipsl.jussieu.fr/orchidee/wiki/GroupActivities/CodeAvailabilityPublication/ORCHIDEE\\_tags\\_2.0\\_gmd\\_2021\\_Africabrowser/tags/ORCHIDEE\\_2\\_1/tags/ORCHIDEE\\_2\\_1](https://forge.ipsl.jussieu.fr/orchidee/wiki/GroupActivities/CodeAvailabilityPublication/ORCHIDEE_tags_2.0_gmd_2021_Africabrowser/tags/ORCHIDEE_2_1/tags/ORCHIDEE_2_1)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Police par défaut

a mis en forme : Police par défaut

a mis en forme : Police par défaut

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

## 8 Author contribution:

G. Marie, S. Luyssaert and P. Peylin designed the experiments and G. Marie carried them out. G. Marie developed the OPENBUGS model code and performed the simulations. G. Marie and S. Luyssaert prepared the manuscript with contributions from all co-authors.

a mis en forme : Par rapport au texte : 0", Numérotation : Recommencer à chaque page

## 9 References

- ALOS - EoPortal Directory - Satellite Missions, [earth.esa.int/web/eoportal/satellite-missions/a/alos\\_tools\\_orchidas.lsce.ipsl.fr/dev/lccci/tools.php](http://earth.esa.int/web/eoportal/satellite-missions/a/alos_tools_orchidas.lsce.ipsl.fr/dev/lccci/tools.php).
- Asner, Gregory P., et al., "A Universal Airborne LiDAR Approach for Tropical Forest Carbon Mapping." *Oecologia*, vol. 168, no. 4, 2011, pp. 1147–1160., doi:10.1007/s00442-011-2165-z.
- Avitabile, Valerio, et al., "An Integrated Pan-Tropical Biomass Map Using Multiple Reference Datasets." *Global Change Biology*, vol. 22, no. 4, 2016, pp. 1406–1420., doi:10.1111/gcb.13139.
- Baccini, A., et al., "Estimated Carbon Dioxide Emissions from Tropical Deforestation Improved by Carbon-Density Maps." *Nature Climate Change*, vol. 2, no. 3, 2012, pp. 182–185., doi:10.1038/nclimate1354.
- Bauer, Peter, et al., "The Quiet Revolution of Numerical Weather Prediction." *Nature*, vol. 525, no. 7567, 2015, pp. 47–55., doi:10.1038/nature14956.
- Beech, E., et al., "GlobalTreeSearch: The First Complete Global Database of Tree Species and Country Distributions." *Journal of Sustainable Forestry*, vol. 36, no. 5, 2017, pp. 454–489., doi:10.1080/10549811.2017.1310049.
- Bonan, Gordon B., et al., "Landscapes as Patches of Plant Functional Types: An Integrating Concept for Climate and Ecosystem Models." *Global Biogeochemical Cycles*, vol. 16, no. 2, 2002, doi:10.1029/2000gb001360.
- Boucher, Olivier, et al., "Presentation and Evaluation of the IPSL-CM6A-LR Climate Model." *Journal of Advances in Modeling Earth Systems*, vol. 12, no. 7, 2020, p. e2019MS002010, doi:<https://doi.org/10.1029/2019MS002010>.
- Bouvet, Alexandre, et al., "An aboveAbove-Ground Biomass Map of African Savannahs and Woodlands at 25 m Resolution Derived from ALOS PALSAR." *Remote Sensing of Environment*, vol. 206, 2018, pp. 156–173., doi:10.1016/j.rse.2017.12.030.
- Brovkin, Victor, et al., "A Continuous Climate-Vegetation Classification for Use in Climate-Biosphere Studies." *Ecological Modelling*, vol. 101, no. 2-3, 1997, pp. 251–261., doi:10.1016/s0304-3800(97)00049-5.
- Buchhorn, Marcel, et al., "Copernicus Global Land Cover Layers—Collection 2." *Remote Sensing*, vol. 12, no. 6, 2020, p. 1044., doi:10.3390/rs12061044.
- Buitenhof, R., et al., "Increased Tree Densities in South African Savannas: >50 Years of Data Suggests CO<sub>2</sub> as a Driver." *Global Change Biology*, vol. 18, no. 2, 2011, pp. 675–684., doi:10.1111/j.1365-2486.2011.02561.x.

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Retrait : Gauche : 0.16", Suspended : 0.2", Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

[“CCI LAND COVER – S2 Prototype Land Cover 20m Map of Africa.” UN.un-spider.org/links-and-resources/data-sources/cci-land-cover-s2-prototype-land-cover-20m-map-africa](#)

a mis en forme : Normal, Retrait : Gauche : 0.16",  
Suspendu : 0.2"

Calzadilla, Alvaro, et al., “Climate Change and Agriculture: Impacts and Adaptation Options in South Africa.” *Water Resources and Economics*, vol. 5, 2014, pp. 24–48., doi:10.1016/j.wre.2014.03.001.

a mis en forme : Couleur de police : Noir

Carreira, Valeria Paula, et al., “Gene-by-Temperature Interactions and Candidate Plasticity Genes for Morphological Traits in *Drosophila Melanogaster*.” *PLoS ONE*, vol. 8, no. 7, 2013, doi:10.1371/journal.pone.0070851.

a mis en forme : Normal, Retrait : Gauche : 0.16",  
Suspendu : 0.2", Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

Chapin, F. Stuart, et al., “Plant Functional Types as Predictors of Transient Responses of Arctic Vegetation to Global Change.” *Journal of Vegetation Science*, vol. 7, no. 3, 1996, pp. 347–58, doi:10.2307/3236278.

a mis en forme : Couleur de police : Noir

Clark, D. B., et al., “The Joint UK Land Environment Simulator (JULES), Model Description – Part 2: Carbon Fluxes and Vegetation Dynamics.” *Geoscientific Model Development*, vol. 4, no. 3, 2011, pp. 701–722., doi:10.5194/gmd-4-701-2011.

a mis en forme : Couleur de police : Noir

Defourny, Pierre, and ESA Land Cover CCI project team. *Dataset Record: ESA Land Cover Climate Change Initiative (Land\_Cover\_cci): Global Land Cover Maps, Version 2.0.7*, 28 Nov. 2019,  
[catalogue.ceda.ac.uk/uuid/b382ebe6679d44b8b0e68ea4ef4b701c](#).

a mis en forme : Couleur de police : Noir

“Development Activities.” *Documentation/Forcings – ORCHIDEE*,  
[forge.ipsl.jussieu.fr/orchidee/wiki/Documentation/Forcings](#).

a mis en forme : Couleur de police : Noir

Dietze, Michael C., et al., “Iterative nearNear-Term Ecological Forecasting: Needs, Opportunities, and Challenges.” *Proceedings of the National Academy of Sciences*, vol. 115, no. 7, 2018, pp. 1424–1432., doi:10.1073/pnas.1710231115.

a mis en forme : Couleur de police : Noir

Druel, A., Peylin, P., Krinner, G., Ciais, P., Viovy, N., Peregón, A., ... & Mironycheva-Tokareva, N. (2017). Towards a more detailed representation of high-latitude vegetation in the global land surface model ORCHIDEE (ORC-HL-VEGv1.0). *Geoscientific Model Development*, 10(12), 4693–4722.

a mis en forme : Couleur de police : Noir, Français (France)

a mis en forme : Couleur de police : Noir

Dubayah, Ralph, et al., “The Global Ecosystem Dynamics Investigation: High-Resolution Laser Ranging of the Earth’s Forests and Topography.” *Science of Remote Sensing*, vol. 1, June 2020, p. 100002, doi:10.1016/j.srs.2020.100002.

a mis en forme : Couleur de police : Noir

Eamus, Derek, et al., “Savannas.” *Vegetation Dynamics*, pp. 383–414., doi:10.1017/cbo9781107286221.017.

a mis en forme : Couleur de police : Noir, Français (France)

Eigentler, L., and J.A. Sherratt. “Spatial Self-Organisation Enables Species Coexistence in a Model for Savanna Ecosystems.” *Journal of Theoretical Biology*, vol. 487, 2020, p. 110122., doi:10.1016/j.jtbi.2019.110122.

a mis en forme : Couleur de police : Noir

Ellison, Aaron M. “Bayesian Inference in Ecology.” *Ecology Letters*, vol. 7, no. 6, 2004, pp. 509–520., doi:10.1111/j.1461-0248.2004.00603.x.

a mis en forme : Couleur de police : Noir, Français (France)

“FAO.org.” *Land Cover Classification System (LCCS)*, Food and Agriculture Organization of the United Nations,

a mis en forme : Couleur de police : Noir, Français (France)

[www.fao.org/land-water/land/land-governance/land-resources-planning-toolbox/category/details/en/c/1036361/](#), Friedlingstein, Pierre, et al., “Global Carbon Budget 2020.” *Earth System Science Data*, vol. 12, no. 4, Dec. 2020, pp. 3269–340, doi:10.5194/essd-12-3269-2020.

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

[GDAL/OGR contributors \(2022\). GDAL/OGR Geospatial Data Abstraction Software Library. Open-Source Geospatial Foundation. URL <https://gdal.org>](https://gdal.org)

Hansen, M. C., et al. "High-Resolution Global Maps of 21st-Century Forest Cover Change." *Science*, vol. 342, no. 6160, 2013, pp. 850–853., doi:10.1126/science.1244693.

Hardin, G. "The Competitive Exclusion Principle." *Science*, vol. 131, no. 3409, 1960, pp. 1292–1297., doi:10.1126/science.131.3409.1292.

Hartley, A. J., MacBean, N., Georgievski, G., & Bontemps, S. (2017). Uncertainty in plant functional type distributions and its impact on land surface models. *Remote Sensing of Environment*, 203, 71-89.

"Home." *ESA Climate Office*, climate.esa.int/en/.

"Home." *ORCHIDEE*, orchidee.ipsl.fr/.

Houghton, R. A., et al. "Carbon Emissions from Land Use and Land-Cover Change." *Biogeosciences*, vol. 9, no. 12, 2012, pp. 5125–5142., doi:10.5194/bg-9-5125-2012.

Huete, Alfredo. "Vegetation's Responses to Climate Variability." *Nature*, vol. 531, no. 7593, Mar. 2016, pp. 181–82, doi:10.1038/nature17301.

Hurt, G. C., et al. "Linking Models and Data on Vegetation Structure." *Journal of Geophysical Research: Biogeosciences*, vol. 115, no. G2, 2010, doi:10.1029/2009jg000937.

"Japan Aerospace Exploration Agency." *JAXA*, global.jaxa.jp/.

Kearsley, Elizabeth, et al. "Conventional Tree Height–Diameter Relationships Significantly Overestimate Aboveground Carbon Stocks in the Central Congo Basin." *Nature Communications*, vol. 4, no. 1, Aug. 2013, p. 2269, doi:10.1038/ncomms3269.

Krinner, G., et al. "A Dynamic Global Vegetation Model for Studies of the Coupled Atmosphere-Biosphere System." *Global Biogeochemical Cycles*, vol. 19, no. 1, 2005, doi:10.1029/2003gb002199.

Le Toan, T., et al. "The BIOMASS Mission: Mapping Global Forest Biomass to Better Understand the Terrestrial Carbon Cycle." *Remote Sensing of Environment*, vol. 115, no. 11, Nov. 2011, pp. 2850–60, doi:10.1016/j.rse.2011.03.020.

Lewis, Simon L., et al. "Increasing Carbon Storage in Intact African Tropical Forests." *Nature*, vol. 457, no. 7232, 2009, pp. 1003–1006., doi:10.1038/nature07771.

Lewis, Sophie C., et al. "Modeling Insights into Deuterium Excess as an Indicator of Water Vapor Source Conditions." *Journal of Geophysical Research: Atmospheres*, vol. 118, no. 2, 2013, pp. 243–262., doi:10.1029/2012jd017804.

Li, Qingyu, et al. "Mapping the Land Cover of Africa at 10 m Resolution from Multi-Source Remote Sensing Data with Google Earth Engine." *Remote Sensing*, vol. 12, no. 4, 2020, p. 602., doi:10.3390/rs12040602.

Li, Wei, et al. "Major Forest Changes and Land Cover Transitions Based on Plant Functional Types Derived from the ESA CCI Land Cover Product." *International Journal of Applied Earth Observation and Geoinformation*, vol. 47, May 2016, pp. 30–39, doi:10.1016/j.jag.2015.12.006.

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Retrait : Gauche : 0.16", Suspendu : 0.2", Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

Li, Wei, et al., "Gross and Net Land Cover Changes in the Main Plant Functional Types Derived from the Annual ESA CCI Land Cover Maps (1992–2015)." *Earth System Science Data*, vol. 10, no. 1, Jan. 2018, pp. 219–34,  
doi:<https://doi.org/10.5194/essd-10-219-2018>.

a mis en forme : Couleur de police : Noir

Liang, Shunlin. *Advances in Land Remote Sensing: System, Modelling, Inversion and Application*. Springer, 2008.

Lurton, Thibaut, et al., "Implementation of the CMIP6 Forcing Data in the IPSL-CM6A-LR Model." *Journal of Advances in Modeling Earth Systems*, vol. 12, no. 4, 2020, p. e2019MS001940,  
doi:<https://doi.org/10.1029/2019MS001940>.

a mis en forme : Couleur de police : Noir

Mermoz, Stéphane, et al., "Decrease of L-Band SAR Backscatter with Biomass of Dense Forests." *Remote Sensing of Environment*, vol. 159, Mar. 2015, pp. 307–17, doi:[10.1016/j.rse.2014.12.019](https://doi.org/10.1016/j.rse.2014.12.019).

a mis en forme : Police par défaut

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

Mills, A. J., et al., "Ecosystem Carbon Storage under Different Land Uses in Three Semi-Arid Shrublands and a Mesic Grassland in South Africa." *South African Journal of Plant and Soil*, vol. 22, no. 3, 2005, pp. 183–190.,  
doi:[10.1080/02571862.2005.10634705](https://doi.org/10.1080/02571862.2005.10634705).

a mis en forme : Couleur de police : Noir

Mills, A. J., et al., "Ecosystem Carbon Storage under Different Land Uses in Three Semi-Arid Shrublands and a Mesic Grassland in South Africa." *South African Journal of Plant and Soil*, vol. 22, no. 3, 2005, pp. 183–190.,  
doi:[10.1080/02571862.2005.10634705](https://doi.org/10.1080/02571862.2005.10634705).

a mis en forme : Couleur de police : Noir

Mitchard, Edward T. A., et al., "Markedly Divergent Estimates of Amazon Forest Carbon Density from Ground Plots and Satellites." *Global Ecology and Biogeography*, vol. 23, no. 8, 2014, pp. 935–946., doi:[10.1111/geb.12168](https://doi.org/10.1111/geb.12168).

a mis en forme : Couleur de police : Noir

Mitchard, Edward Ta, et al., "Uncertainty in the Spatial Distribution of Tropical Forest Biomass: A Comparison of Pan-Tropical Maps." *Carbon Balance and Management*, vol. 8, no. 1, 2013, doi:[10.1186/1750-0680-8-10](https://doi.org/10.1186/1750-0680-8-10).

a mis en forme : Couleur de police : Noir

Naidoo, Laven, et al., "Savannah Woody Structure Modelling and Mapping Using Multi-Frequency (X-, C- and L-Band) Synthetic Aperture Radar Data." *ISPRS Journal of Photogrammetry and Remote Sensing*, vol. 105, July 2015, pp. 234–50, doi:[10.1016/j.isprsjprs.2015.04.007](https://doi.org/10.1016/j.isprsjprs.2015.04.007).

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

NiSAR : First global SAR mission, in partnership with ISRO, <https://nisar.jpl.nasa.gov>.

a mis en forme : Normal, Retrait : Gauche : 0.16", Suspendu : 0.2", Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir, Français (France)

a mis en forme : Français (France)

a mis en forme : Couleur de police : Noir, Français (France)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

O'Connor, Tim G, et al., "Bush Encroachment in Southern Africa: Changes and Causes." *African Journal of Range & Forage Science*, vol. 31, no. 2, 2014, pp. 67–88., doi:[10.2989/10220119.2014.939996](https://doi.org/10.2989/10220119.2014.939996).

a mis en forme : Couleur de police : Noir, Français (France)

a mis en forme : Français (France)

Oke, T. R. *Boundary Layer Climates*. Routledge, 2002.

Organisation Des Nations Unies Pour L'alimentation Et L'agriculture. *Food and Agriculture Organization of the United Nations*, [www.fao.org/home/fr/](http://www.fao.org/home/fr/).

a mis en forme : Couleur de police : Noir, Français (France)

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

Palmer, Paul I., et al., "Net Carbon Emissions from African Biosphere Dominate Pan-Tropical Atmospheric CO<sub>2</sub> Signal." *Nature Communications*, vol. 10, no. 1, 2019, doi:[10.1038/s41467-019-11097-w](https://doi.org/10.1038/s41467-019-11097-w).

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

Poulter, B., et al., "Plant Functional Type Mapping for Earth System Models." *Geoscientific Model Development*, vol. 4, no. 4, 2011, pp. 993–1010., doi:[10.5194/gmd-4-993-2011](https://doi.org/10.5194/gmd-4-993-2011).

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure), Taquets de tabulation : 3.13", Centré + 6.27", Droite

- Quegan, Shaun, et al.: "The European Space Agency BIOMASS Mission: Measuring Forest ~~above~~Above-Ground Biomass from Space." *Remote Sensing of Environment*, vol. 227, June 2019, pp. 44–60, doi:10.1016/j.rse.2019.03.032.
- Quéré, C. Le, et al.: "Global Carbon Budget 2015." *Earth System Science Data*, vol. 7, no. 2, 2015, pp. 349–396., doi:10.5194/essd-7-349-2015.
- Réjou-Méchain, Maxime, et al.: "Using Repeated Small-Footprint LiDAR Acquisitions to Infer Spatial and Temporal Variations of a High-Biomass Neotropical Forest." *Remote Sensing of Environment*, vol. 169, 2015, pp. 93–101., doi:10.1016/j.rse.2015.08.001.
- Rietkerk, M., Boerlijst, M. C., van Langeveld, F., HilleRisLambers, R., de Koppel, J. V., Kumar, L., ... & de Roos, A. M. (2002). Self-organization of vegetation in arid ecosystems. *The American Naturalist*, 160(4), 524–530.
- Saatchi, S. S., et al.: "Benchmark Map of Forest Carbon Stocks in Tropical Regions across Three Continents." *Proceedings of the National Academy of Sciences*, vol. 108, no. 24, 2011, pp. 9899–9904., doi:10.1073/pnas.1019576108.
- Sankaran, Mahesh, et al.: "Determinants of Woody Cover in African Savannas." *Nature*, vol. 438, no. 7069, 2005, pp. 846–849., doi:10.1038/nature04070.
- Santoro, Maurizio, et al.: "The Global Forest Above-Ground Biomass Pool for 2010 Estimated from High-Resolution Satellite Observations." *Earth System Science Data Discussions*, July 2020 August 2021, pp. 1–38, doi:10.5194/essd-2020-148,13-3927-2021.
- Schaaf: MODIS Albedo and Reflectance Anisotropy Products. Accessed 24 Mar. 2021.
- Shimada, Masanobu, and Takahiro Ohtaki. "Generating Large-Scale High-Quality SAR Mosaic Datasets: Application to PALSAR Data for Global Monitoring." *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, vol. 3, no. 4, 2010, pp. 637–656., doi:10.1109/jstars.2010.2077619.
- Simard, Marc, et al.: "Mapping Forest Canopy Height Globally with Spaceborne Lidar." *Journal of Geophysical Research: Biogeosciences*, vol. 116, no. G4, 2011, doi:<https://doi.org/10.1029/2011JG001708>.
- Sitch, S., et al.: "Evaluation of Ecosystem Dynamics, Plant Geography and Terrestrial Carbon Cycling in the LPJ Dynamic Global Vegetation Model." *Global Change Biology*, vol. 9, no. 2, 2003, pp. 161–185., doi:10.1046/j.1365-2486.2003.00569.x.
- Still, Christopher J., et al.: "Global Distribution of C3 and C4 Vegetation: Carbon Cycle Implications." *Global Biogeochemical Cycles*, vol. 17, no. 1, 2003, pp. 6-1-6-14, doi:<https://doi.org/10.1029/2001GB001807>.
- Sun, Wanxiao, et al.: "Mapping Plant Functional Types from MODIS Data Using Multisource Evidential Reasoning." *Remote Sensing of Environment*, vol. 112, no. 3, Mar. 2008, pp. 1010–24, doi:10.1016/j.rse.2007.07.022.
- Thomas, Neal, 2010. "Overview". [OpenBUGS website](#). Retrieved 9 October 2010.
- Tyukavina, Alexandra, et al.: "Congo Basin Forest Loss Dominated by Increasing Smallholder Clearing." *Science Advances*, vol. 4, no. 11, 2018, doi:10.1126/sciadv.aat2993.

UN-spider: "CCI LAND COVER - S2 Prototype Land Cover 20m Map of Africa." [UN.un-spider.org/links-and-resources/data-sources/ccj-land-cover-s2-prototype-land-cover-20m-map-africa](http://UN.un-spider.org/links-and-resources/data-sources/ccj-land-cover-s2-prototype-land-cover-20m-map-africa).

a mis en forme : Normal, Retrait : Gauche : 0.16",  
Suspendu : 0.2"

Viovy, Nicolas, 2017, CRUNCEP data set:

<https://vesg.ipsl.upmc.fr/thredds/fileServer/work/p529viov/cruncep/readme.html>

Wigley, B. J., et al. "Bush Encroachment under Three Contrasting Land-Use Practices in a Mesic South African Savanna." *African Journal of Ecology*, vol. 47, 2009, pp. 62–70., doi:10.1111/j.1365-2028.2008.01051.x.

a mis en forme : Couleur de police : Noir

a mis en forme : Normal, Retrait : Gauche : 0.16",  
Suspendu : 0.2", Bordure : Haut: (Pas de bordure), Bas: (Pas de bordure), Gauche: (Pas de bordure), Droite: (Pas de bordure), Entre : (Pas de bordure)

a mis en forme : Couleur de police : Noir