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Supporting Information for

Uncertainty in Land Carbon Fluxes Simulated by CMIP6 Models from Inconsistent Treatments of Crop Distributions and Photosynthetic Pathways

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Supplementary:**Material and Method S1 - Description of implementation of vegetation cover in CMIP6 models****S1.1 ACCESS-ESM1.5**

In the Australian Community Climate and Earth System Simulator Earth System Model version 1.5 (ACCESS-ESM1.5) land surface and biogeochemical processes are simulated using the Community Atmosphere Biosphere Land Exchange (CABLE) (Kowalczyk et al., 2013) model and a modified Carnegie–Ames–Stanford Approach (CASA) carbon (C) cycle model which incorporates nitrogen (N) and phosphorus (P) cycles (CASA-CNP module) (Ziehn et al., 2020). The area abundance of crops for the historical period is based on the LUH2 dataset (Hurtt et al., 2020). The 12 land-use types in the LUH2 dataset are mapped into 9 plant functional types (PFTs) in CABLE including C₃ grass, C₃ crops and C₄ crops. The rest of the PFTs (evergreen needleleaf forest, evergreen broadleaf forest, deciduous needleleaf forest, deciduous broadleaf forest, shrub and tundra) are not specified as either C₃ or C₄ but most of them would be C₃ vegetation since it is the most prevalent of these two photosynthetic pathways in the given PFTs. However, even though a C₄ crop PFT was specified in the model description (Ziehn et al., 2020), the variable cropFracC₄ was not provided in the CMIP6 archive. The values for cropFracC₃ and for the total crop fraction (cropFrac) were identical, so we specified the fraction of C₄ crops to be zero everywhere.

S1.2 CanESM5

The land component of the Canadian Earth System Model version 5 (CanESM5) is comprised of the Canadian Land Surface Scheme (CLASS) and the Canadian Terrestrial Ecosystem Model (CTEM) making up the CLASS-CTEM modelling framework. CLASS simulates the physical and CTEM simulates the biogeochemical land surface processes. Energy and water balances and land surface physical processes are calculated in CLASS using four vegetation categories: needleleaf trees, broadleaf trees, grasses and crops. Each of these is divided into sub-categories: needleleaf trees into evergreen and deciduous, broadleaf trees into three categories; cold deciduous, drought deciduous, and evergreen, crops into C₃ and C₄ types, and grasses into C₃ and C₄ types to form 9 PFTs (Swart et al., 2019). The increase in the area abundance of crops s for the historical period is based on the LUH2 dataset (Hurtt et al., 2020).

S1.3 CESM2, CESM2-WACCM, CMCC-CM2-SR5 and CMCC-ESM2-HAM

The Community Earth System Model Version 2 (CESM2) and Whole Atmosphere Community Climate Model (WACCM) are developed by teams of individuals from the National Centre for Atmospheric Research (NCAR), and other research institutions and universities. The major difference between CESM2 and CESM2-WACCM is the vertical extent of their atmospheric configuration. In CESM2-WACCM the simulated properties extend up to 130-140 km in the upper atmosphere with 70 vertical levels (Gettelman et al., 2019) whereas CESM2 has only 32 vertical levels and a model top of 40 km, without a prognostic scheme for ozone and other stratospheric entities (Danabasoglu et al., 2020).

CESM2 and CESM2-WACCM use the Community Land Model Version 5 (CLM5) for the simulation of terrestrial biogeochemical cycles including interactions with the soil-plant-atmospheric continuum, anthropogenic land use change and vegetation coverage (Danabasoglu et al., 2020; Lawrence et al., 2019). CLM5 computes land unit weights through two methods: dataset input or prognostic representation. In the prognostic representation, the land surface heterogeneity is represented in several land units which are divided into columns and the columns are divided into patches. The land units are vegetated land, lake, urban, glacier, and cropland. The vegetated land unit is divided into patches of 14 plant or crop functional type (P/CFTs) which are (N = needleleaf; B = broadleaf; E = evergreen; D = deciduous; T = tree; S = shrub): NET temperate, NET boreal, NDT boreal, BET tropical, BET temperate, BDT tropical, BDT temperate, BDT boreal, BES temperate, BDS temperate, BDS boreal, C₃ Arctic grass, C₃ grass and C₄ grass. Crops are categorised into C₃ unmanaged rainfed crop, C₃ unmanaged irrigated crop, managed rainfed unirrigated crops and managed irrigated crops. Apart from the last three PFTs, it is not explicitly stated which of the other PFTs uses the C₃ or C₄ photosynthetic pathways. However, their output variables provided include the percentage area coverage of all C₃ and C₄ vegetation. The CLM uses its Land Use Data Tool to translate the LUH2 dataset transitions and information into its P/CFTs distribution and management (Hurtt et al., 2020; Lawrence et al., 2019). The

CESM2 models did not provide cropFracC3 and cropFracC4 variables, but they did provide c3PftFrac or c4PftFrac (the fractions for total vegetation), so cropFracC3 and cropFracC4 were calculated by multiplying cropFrac by c3PftFrac or c4PftFrac.

The Euro-Mediterranean Centre on Climate Change (CMCC) Foundation model simulates biogeochemical cycles and land use change with the Community Land Model version 4.5 (CLM4.5) (Lovato et al., 2022), which is an earlier version of the model used in CESM2 and CESM2-WACCM. The land use implementation is similar to CESM2, which implies they used the LUH2 dataset as well (Lawrence et al., 2019). However, unlike CESM2, the CMCC models do provide cropFracC3, although not cropFracC4. Their cropFracC3 was equal to the total crop fraction (cropFrac), so we specified the fraction of C4 crops to be zero everywhere

S1.4 CNRM-CM6.1 and CNRM-ESM2.1

CNRM-CM6.1 is the sixth generation of the fully coupled atmosphere-ocean general circulation model jointly developed by Centre National de Recherches Météorologiques (CNRM) and Cerfacs. Its land surface is simulated with the Interaction Soil-Biosphere-Atmosphere-CNRM TRIP (ISBA-CTRIP) coupled system (Decharme et al., 2019). Its vegetation cover is based on the ECOCLIMAP-II database in which each land cover type is portioned into fractions of 4 main surface types or tiles (nature, water bodies, sea, urban areas). The nature tile is further divided into bare soil, bare rock or one of ten PFTs. CNRM-CM6-1 adopts a fixed vegetation cover map from ECOCLIMAP-II (Faroux et al., 2013; Voldoire et al., 2019). It does not incorporate LUH2 datasets.

CNRM-ESM2.1 and CNRM-CM6.1 share the same code, physical parameterization and grid resolution but vegetation coverage in CNRM-ESM2.1 is dynamic, incorporating the LUH2 crop coverage dataset. In CNRM-CM6-1 and CNRM-ESM2.1, the cropFracC3 and cropFracC4 variables were identical to the c3PftFrac and c4PftFrac variables, which appears to be an error in the output. So, in CNRM-CM6-1 and CNRM-ESM2.1 cropFracC3 and cropFracC4 were calculated by multiplying cropFrac by c3PftFrac or c4PftFrac.

S1.5 MPI-ESM1.2

The Max Planck Institute Earth System Model version 1.2 (MPI-ESM1.2) uses the land surface scheme JSBACH which focuses on the competition between two vegetation types: graminoids (grass-like) and woody species (trees and shrubs). These two broad categories are further divided into different PFTs. To integrate the LUH2 dataset into their JSBACH simulation, they aggregated the four LUH2 natural vegetation and grasslands states into the natural vegetation type, the five LUH2 crop variables into crops, the LUH2 pasture and urban into pasture. These three broad types: natural vegetation, crops and pastures are then proportionally adjusted to sum up to one (Mauritsen et al., 2019). JSBACH dynamic vegetation component DYNVEG has 8 PFTs, 6 of which are woody (tropical evergreen trees, tropical deciduous trees, extratropical evergreen trees, extratropical deciduous trees, raingreen shrubs, deciduous shrubs) and 2 of which are grass- (C3 grasses and C4 grasses). In the model, prevalent climate conditions determine the extent of the presence or absence of vegetation in hot and cold regions (Reick et al., 2013).

S1.6 UKESM1-0-LL

The UK Earth System Model version 1 (UKESM1) (Mulcahy et al., 2023; Sellar et al., 2019) uses a physical core built on a prior model, HadGEM3-GC3.1 (Kuhlbrodt et al., 2018; Williams et al., 2018). It simulates the terrestrial ecosystem biogeochemistry and land surface dynamics using the Joint UK Land Environment Simulator (JULES) (Clark et al., 2011; Littleton et al., 2020) which is built on the Top-down Representation of Interactive Foliage and Flora Including Dynamics (TRIFFID) which uses the LUH2 crop cover data with some modifications including simulation of competition between C3 and C4 (Cox et al., 2001). ~~It uses, and the~~ Rothamsted carbon model soil carbon scheme (RothC) (Falloon et al., 1998). TRIFFID is a dynamic vegetation model that divides the earth surface into five components: vegetated land, urban and lake surface types, bare soil, and land covered permanently by ice. The vegetated component is divided into five PFTs : broadleaf trees, needleleaf trees, C3 (temperate) grasses, C4 (tropical) grasses and shrubs. These are assumed to sufficiently represent the phenotypic and genotypic variations needed to capture biophysical and biogeochemical roles of vegetation in earth system models (Sellar et al., 2019, Clark et al., 2011).

Table S1. Global land area in each category in LUH2 and in CMIP6 models for 1970 and 2014.

	Area Abundance (Million km ²)			
	C3 crop	C4 Crop	C3 Natural	C4 Natural
LUH2 1970 (2014)	11.59 (12.75)	2.68 (3.34)		
Luo (2014)		(3.32)		(17.02)
Ensemble mean 1970 (2014)	10.71 (11.55)	1.81 (2.31)	66.76 (66.41)	15.33 (15.29)
ACCESS-ESM1-5 1970 (2014)	12.41 (13.76)	0.00 (0.00)	67.66 (67.35)	9.97 (8.95)
CanESM5 1970 (2014)	11.43 (12.41)	2.68 (3.30)	64.96 (63.92)	6.88 (6.33)
CESM2 1970 (2014)	9.42 (9.94)	3.04 (3.92)	71.34 (70.48)	19.44 (18.91)
CESM2-WACCM 1970 (2014)	9.42 (9.94)	3.04 (3.92)	71.34 (70.48)	19.44 (18.91)
CMCC-CM2-SR5 1970 (2014)	13.21 (14.68)	0.00 (0.00)	68.99 (67.79)	19.63 (19.35)
CMCC-ESM2 1970 (2014)	13.21 (14.68)	0.00 (0.00)	68.99 (67.79)	19.63 (19.35)
CNRM-CM6.1 Fixed	9.71 (9.71)	3.00 (3.00)	59.82 (59.82)	19.22 (19.22)
CNRM-ESM2.1 1970 (2014)	9.48 (10.03)	2.26 (3.10)	65.01 (65.37)	20.09 (19.56)
MPI-ESM1.2-HAM 1970 (2014)	11.89 (13.19)	1.35 (1.96)	70.72 (71.18)	8.75 (10.18)
MPI-ESM1-2-LR 1970 (2014)	11.96 (13.03)	1.14 (1.60)	68.33 (68.59)	8.79 (9.91)
UKESM1-LL 1970 (2014)	5.68 (5.65)	3.36 (4.62)	57.12 (57.76)	16.84 (17.47)

Table S2. Ranges of simulated terrestrial biosphere carbon flux variables for 2014 simulated by CMIP6 models from 1850 to 2014.. $\div$ GPP is Gross Primary Production; C_{veg} is vegetation carbon content; dC_{veg}/dt is vegetation carbon accumulation; C_{land} is land carbon content; dC_{land}/dt is land carbon accumulation .

Variable	Area (Million km ²)	GPP (PgCyr ⁻¹)	Discrimination (‰)	Discrimination change (‰)
Total	95 [85, 105]	116 [93, 150]	18.12 [16.98, 19.15]	-0.23 [-0.68, 0.20]
Total Natural	82 [71, 90]	100 [80, 126]	18.15 [17, 19.42]	-0.17 [-0.80, 0.22]
Total Crop	14 [10, 15.4]	17 [13, 26]	-0.06 [-0.7, 0.33]	-0.07 [-0.50, 0.20]
C ₃ total	78 [63, 84]	96 [73, 126]		
C ₃ Natural	66 [57, 71]	82 [62, 105]		
C ₃ crop	12 [6, 15]	14 [6, 21]		
C ₄ total	18 [9, 13]	21 [12, 30]		
C ₄ Natural	15 [6, 19]	18 [15, 26]		
C ₄ crop	2 [0, 4.6]	3 [0, 8]		

Table S2 biosphere carbon accumulation.

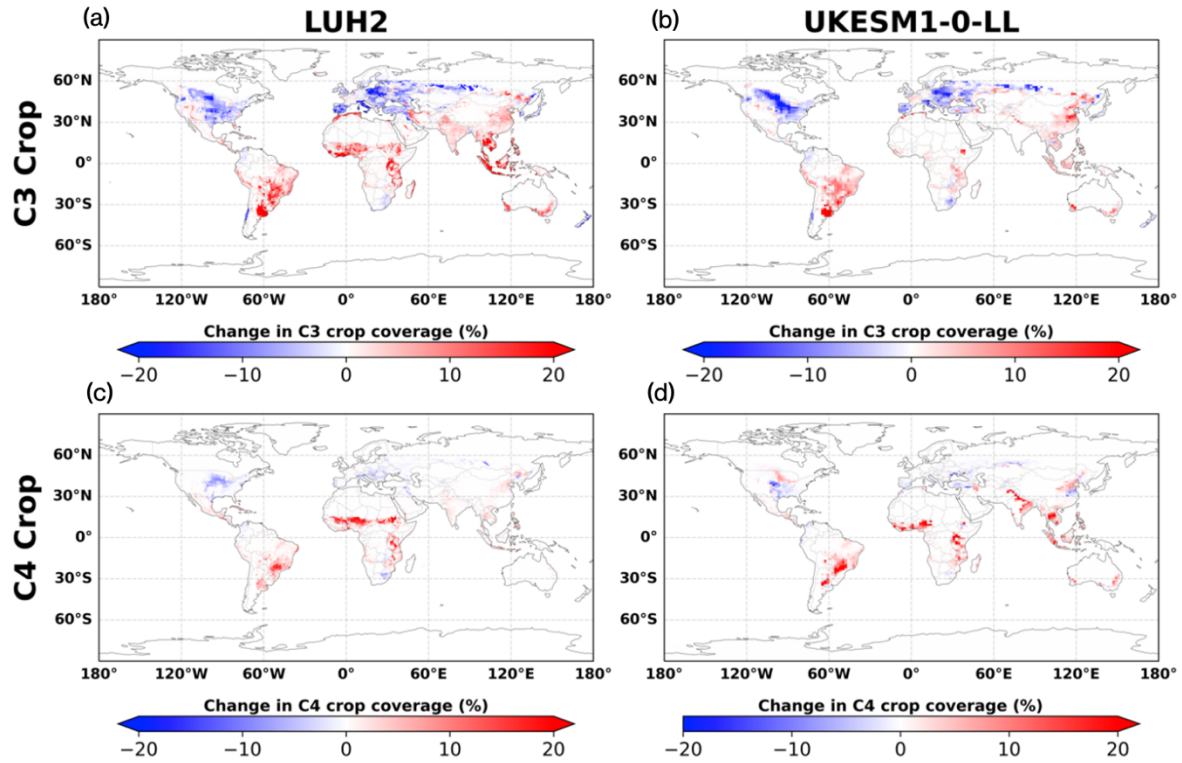


Figure S1. Percentage change in crop abundance between 1970 and 2014. For C₃ and C₄ in LUH2 (a) and (c) and UKESM1-0-LL (b) and (d). All the other models not shown use the LUH2 crop variables for their crop coverage and change. CNRM-CM6-1 is not shown here because it has fixed crop coverage.

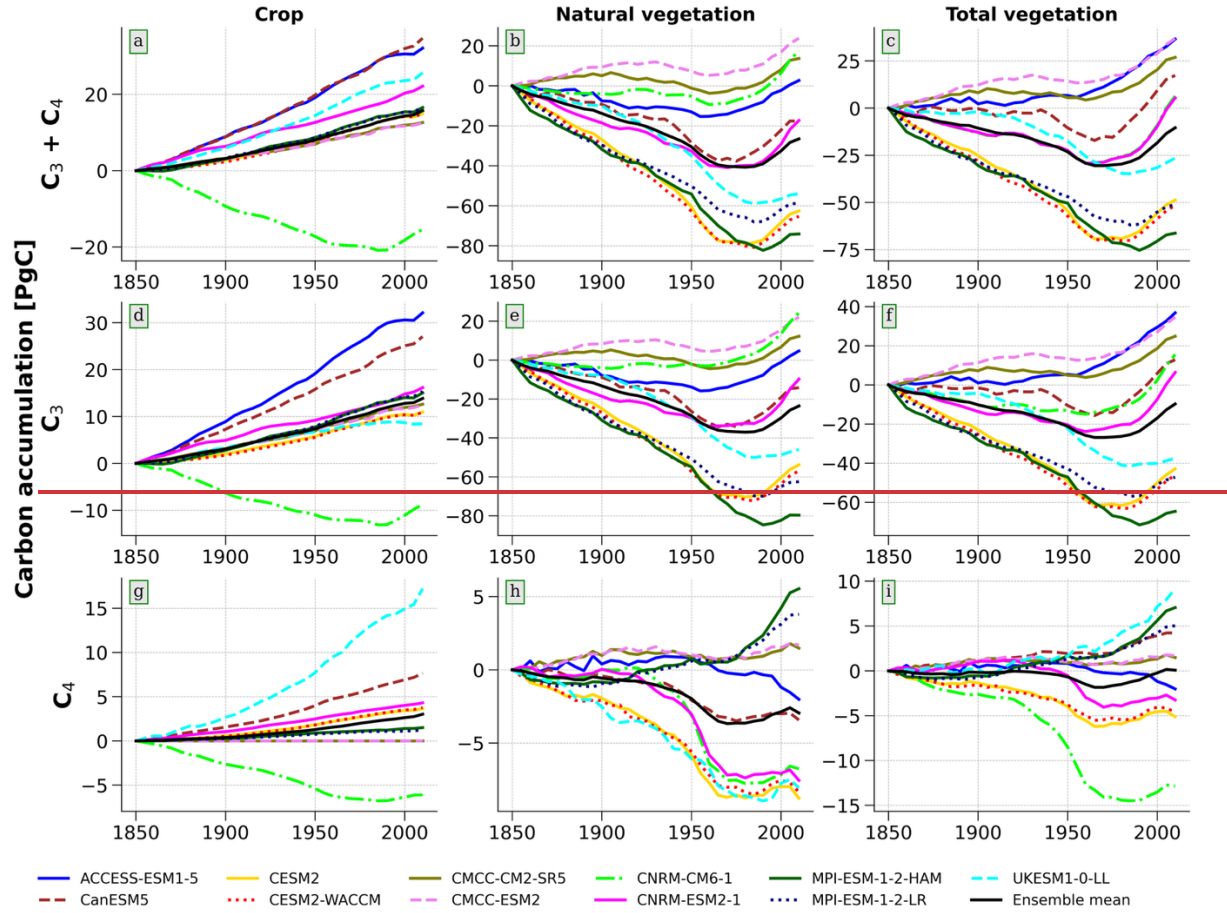


Figure S2—Temporal trend in gGlobal vegetation carbon accumulation between 1850 and 2014. Values of carbon accumulation in (a) croplands, (b) natural vegetation and (c) total vegetation. (d) C_3 -crop, (e) natural C_3 vegetation, (f) total C_3 vegetation, (g) C_4 -crops, (h) natural C_4 vegetation and (i) total C_4 vegetation

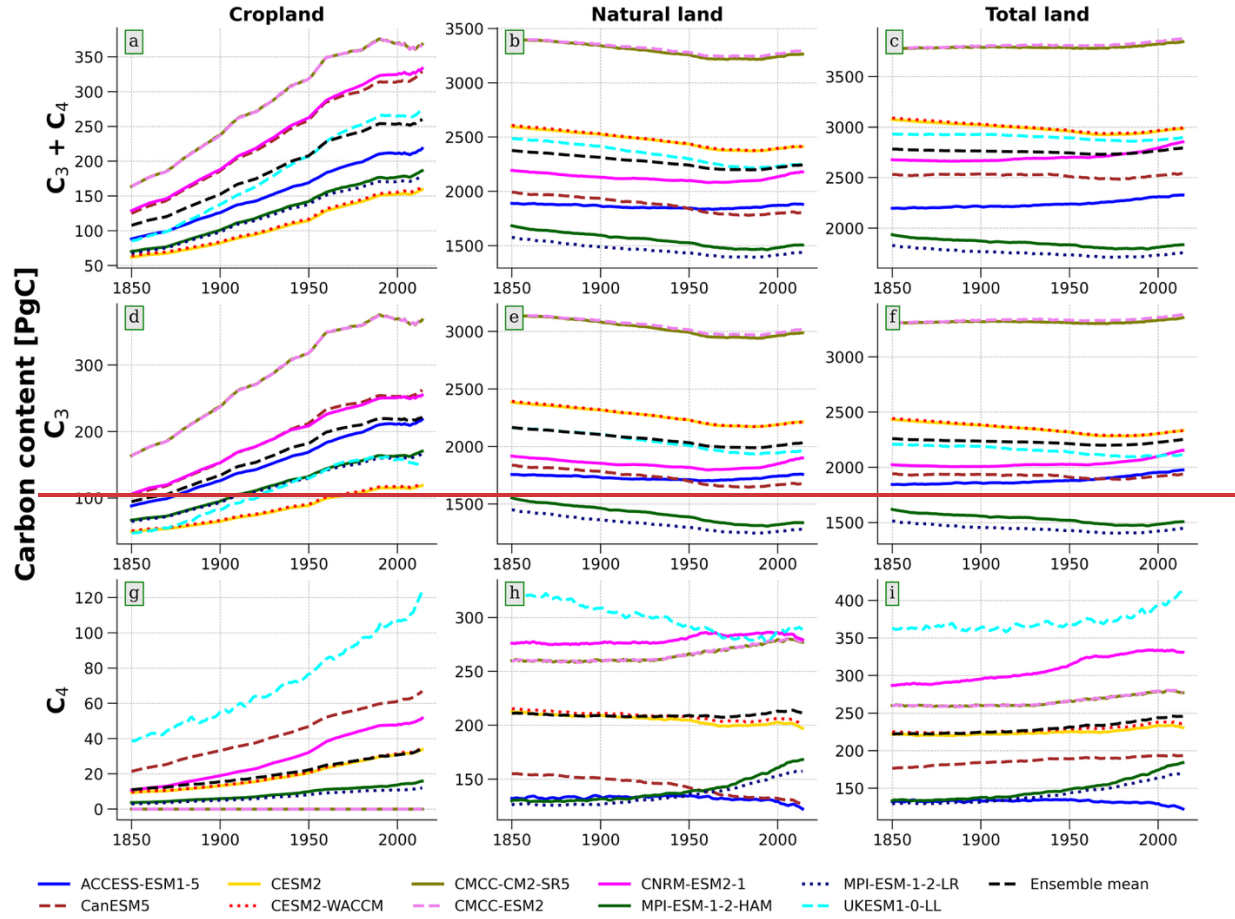


Figure S3—Temporal trend in gGlobal land carbon content between 1850 and 2014. Absolute values of carbon content in (a) croplands, (b) natural vegetated land (c) total vegetated land, (d) C_3 croplands, (e) natural C_3 vegetated land, (f) total C_3 vegetated land, (g) C_4 croplands, (h) natural C_4 vegetated land and (i) total C_4 vegetated lands.

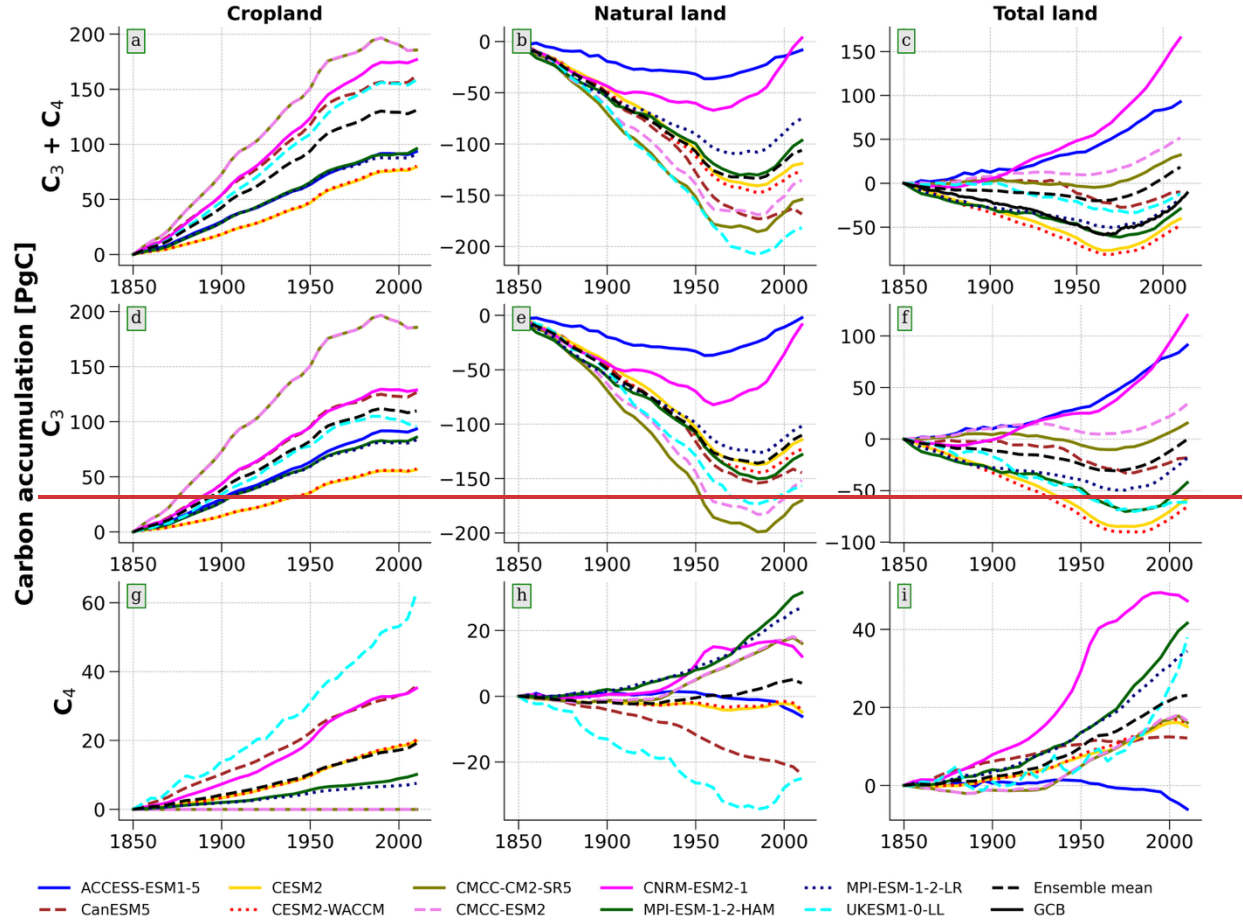


Figure S4—Temporal trend in gGlobal land carbon accumulation between 1850 and 2014. Absolute values of carbon content in (a) croplands, (b) natural vegetated land (c) total vegetated land, (d) C_3 croplands, (e) natural C_3 vegetated land, (f) total C_3 vegetated land, (g) C_4 croplands, (h) natural C_4 vegetated land and (i) total C_4 vegetated lands. Estimated global land carbon accumulation from the Global Carbon Budget

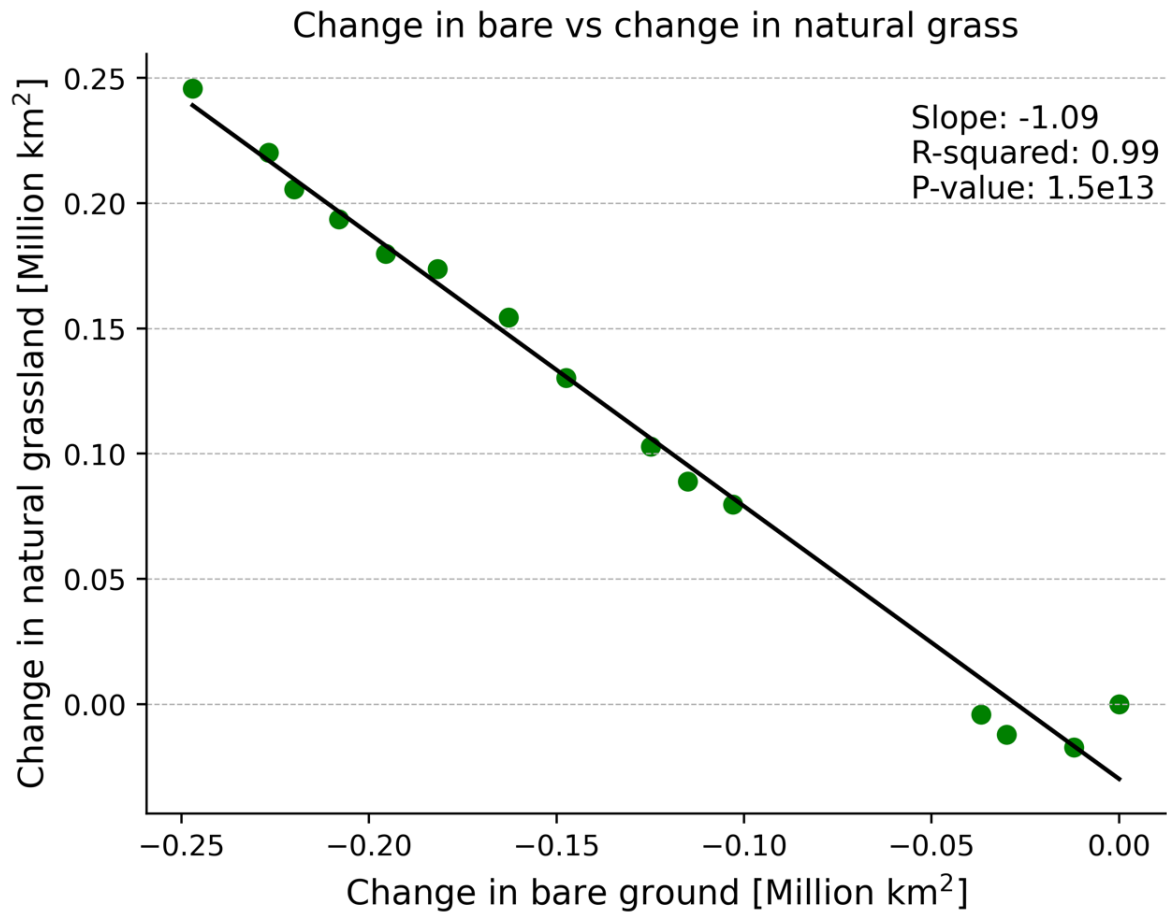


Figure S25 – Relationship between the change in bare ground and natural grass in the ESA CCI land cover dataset between 2000 and 2014. The change is calculated as the difference between the values of the variables in 2000 and 2014.

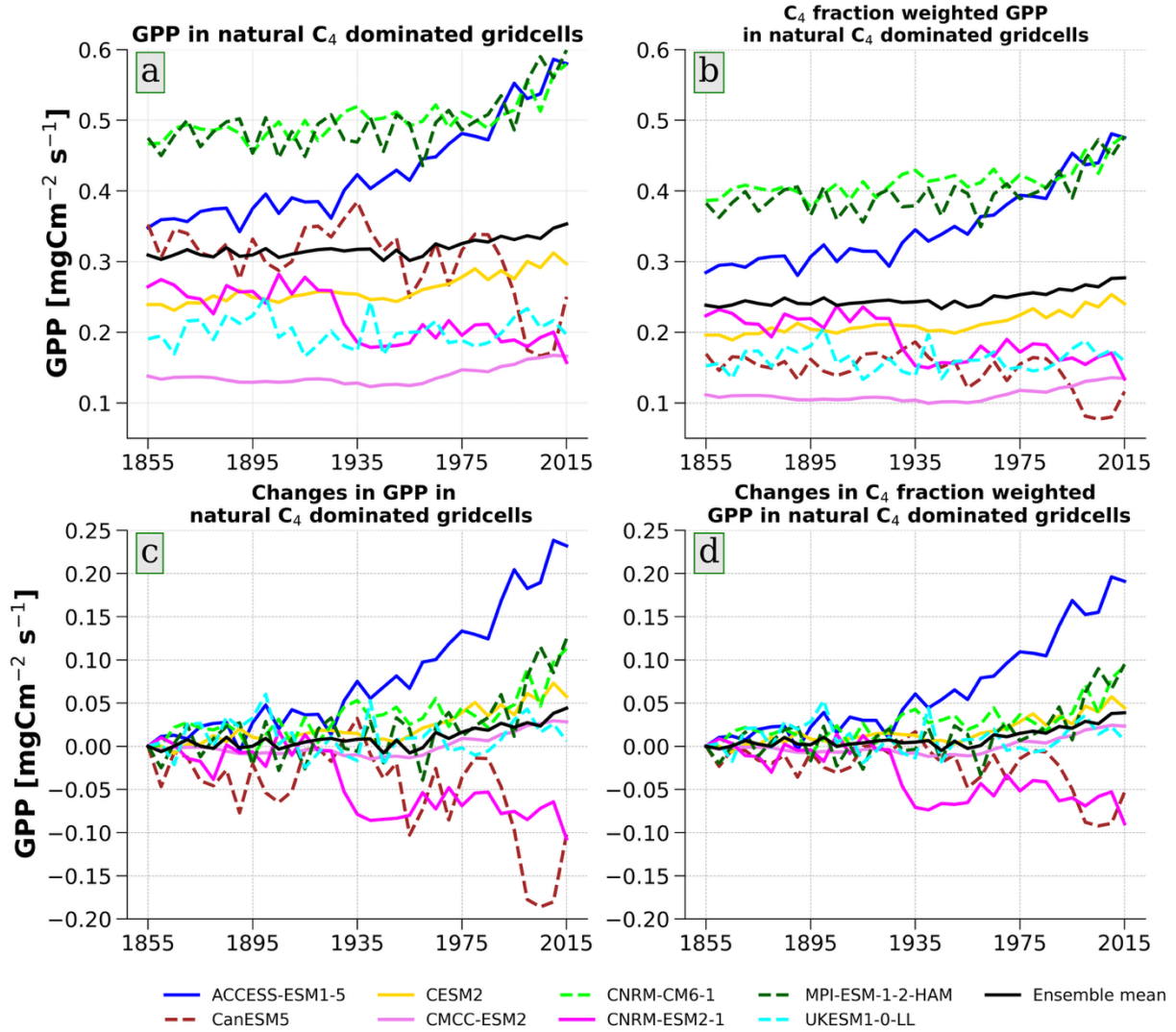


Figure S36 - Temporal trend of GPP in natural C₄ dominated grid cells between 1850 and 2014. Here natural C₄ $\Rightarrow$ 75% except in CanESM5 where there are no grid cells with natural C₄ $\Rightarrow$ 75% and as such a threshold of natural C₄ $\Rightarrow$ 45.5% is used instead. (a) GPP in natural C₄ dominated grid cells, (b) GPP in natural C₄ dominated grid cells weighted by the natural C₄ vegetation fraction, (c) Changes in the GPP in natural C₄ dominated grid cells, (d) changes in the GPP in natural C₄ dominated grid cells weighted by the natural C₄ vegetation fraction.

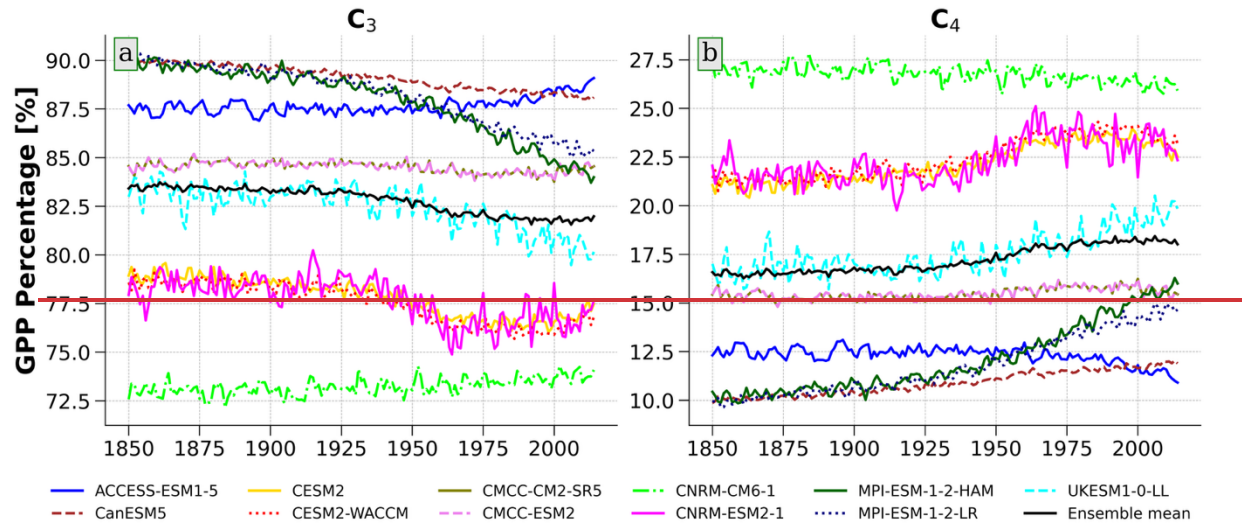


Figure S7— Temporal trend of C_3 and C_4 GPP percentage in CMIP6 models between 1850 and 2014. The model means show a decreasing C_3 and an increasing C_4 contribution to the total GPP.

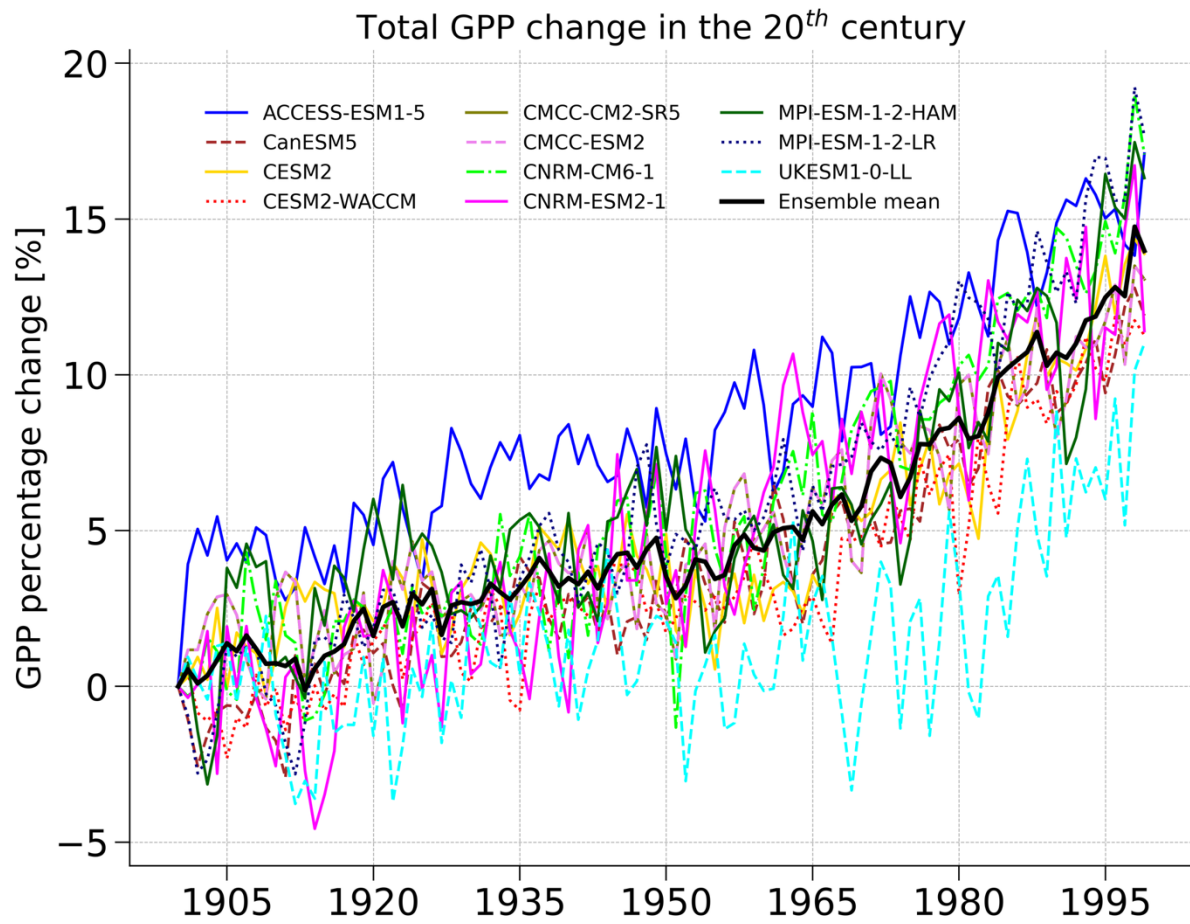


Figure S48– CMIP6 models percentage change in total GPP between 1900 and 2000. The value of the percentage change in the GPP is weak (11 to 17%) compared to 30% observed in other studies in the 20th century.

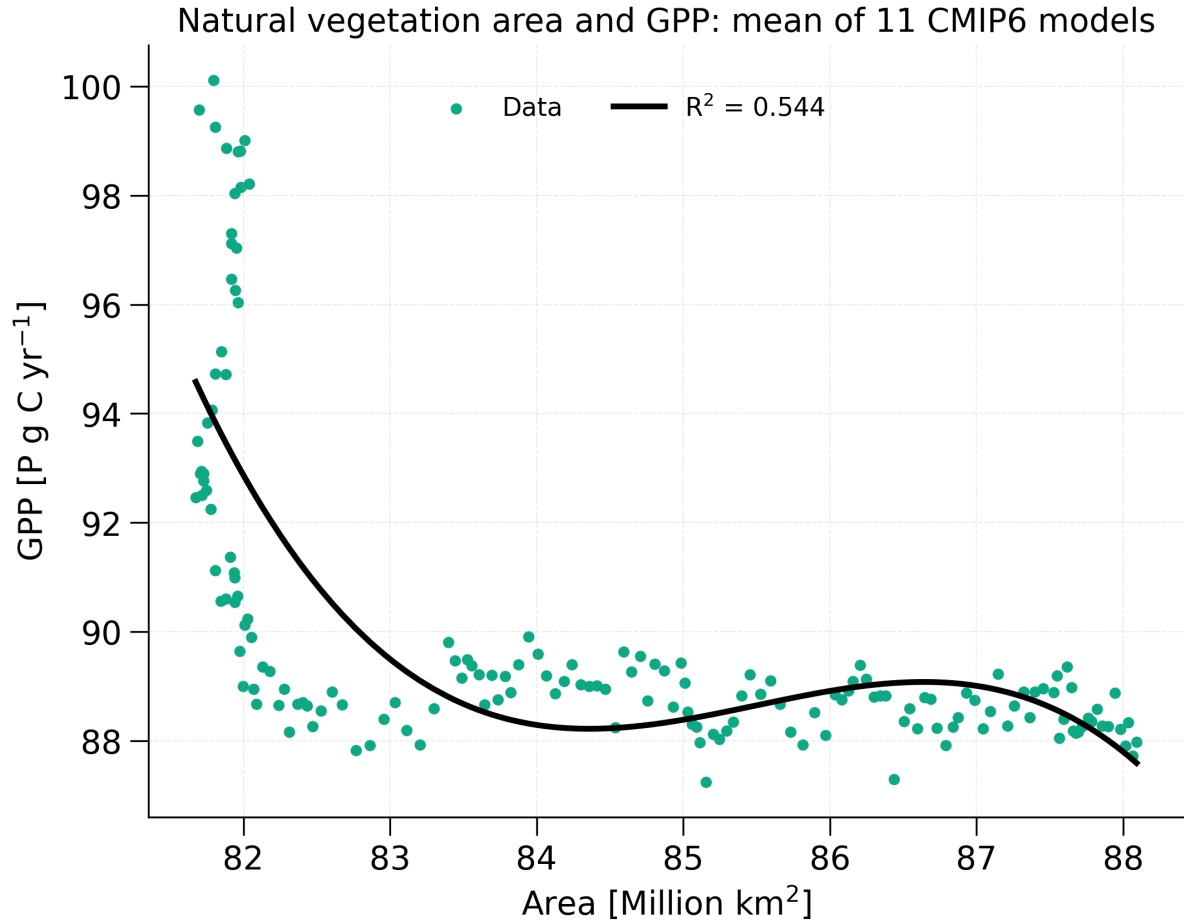


Figure S59— Relationship between GPP and area for natural vegetation for the mean of the 11 CMIP6 models. The line is fitted with a cubic polynomial and $R^2 = 0.544$. It shows GPP is decreasing with increasing area which is contrary to what is theoretically expected which is a decrease in GPP with decreasing area, hence our assumption that the change in GPP in natural vegetation may be decoupled from the change in its area.

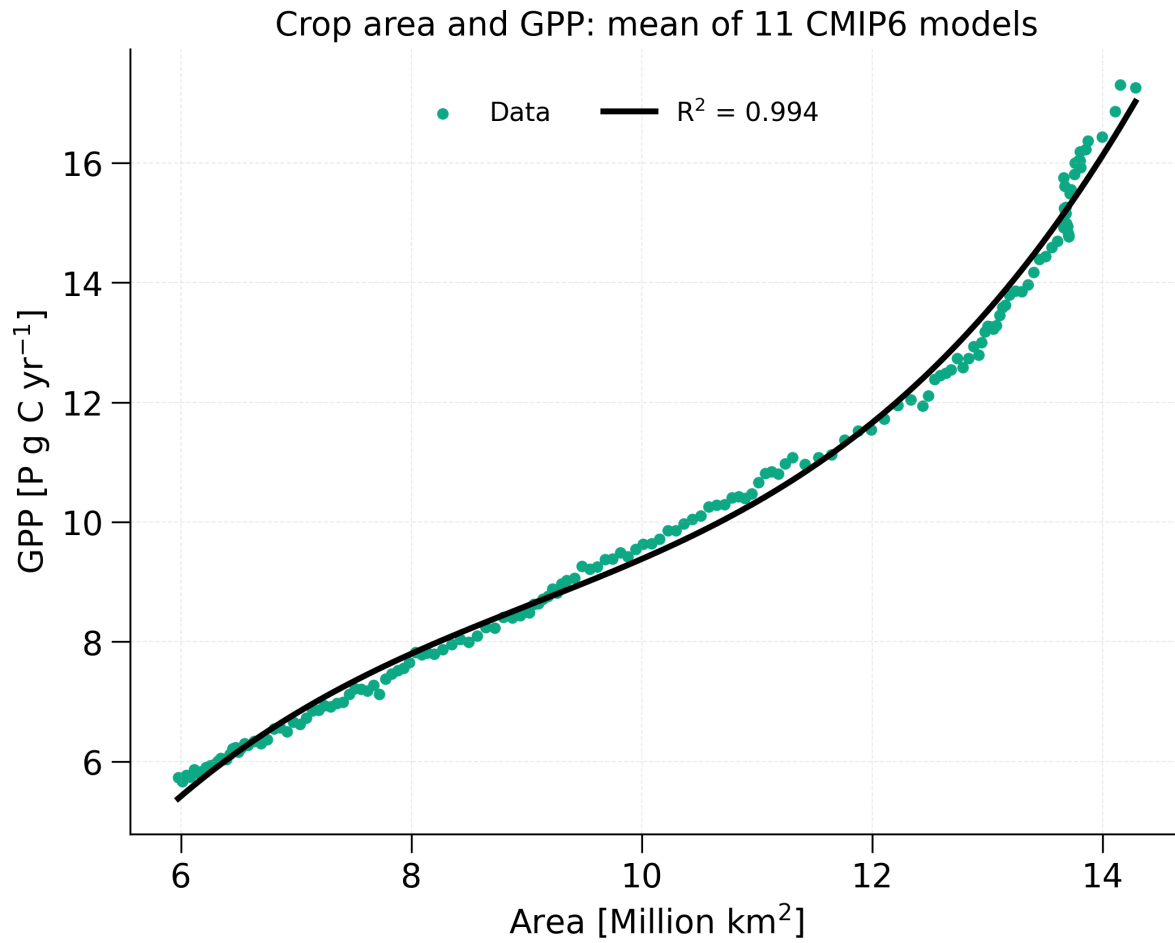


Figure S610— Relationship between GPP and area for cropland for the mean of the 11 CMIP6 models. The line is fitted with a cubic polynomial and $R^2 = 0.994$. The GPP increased with the increasing area.

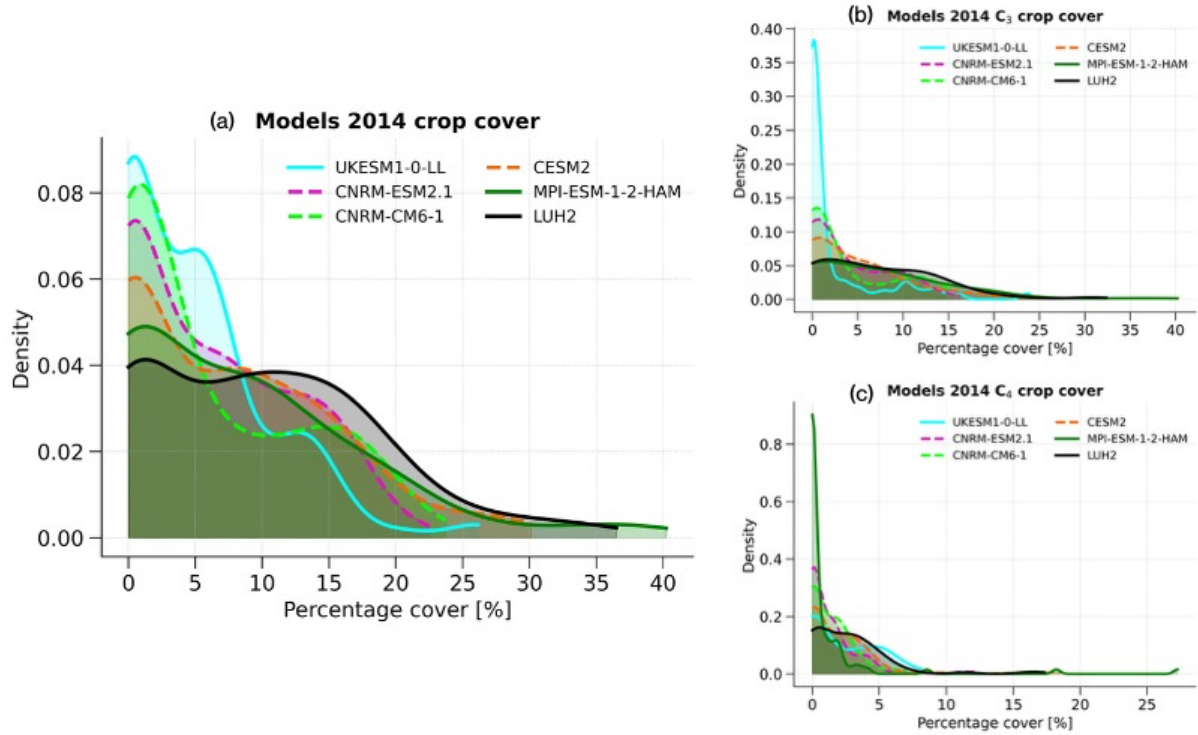


Figure S744 – Distribution density plot for CMIP6 models and LUH2 crop cover for 2014. Total crop (a), C₃ crop (b), and C₄ crop (c). The two models which have outliers for C₃ and C₄ crops are UKESM1 and MPI-ESM-1-2-HAM. The values we obtained by multiplying the total crop cover and the total C₃ and C₄ fraction to get the values of C₃ and C₄ crops for CESM2, CNRM-CM6.1 and CNRM-ESM2.1 are within the range of the other models and they are comparable to the values of the LUH2 values and the other models.

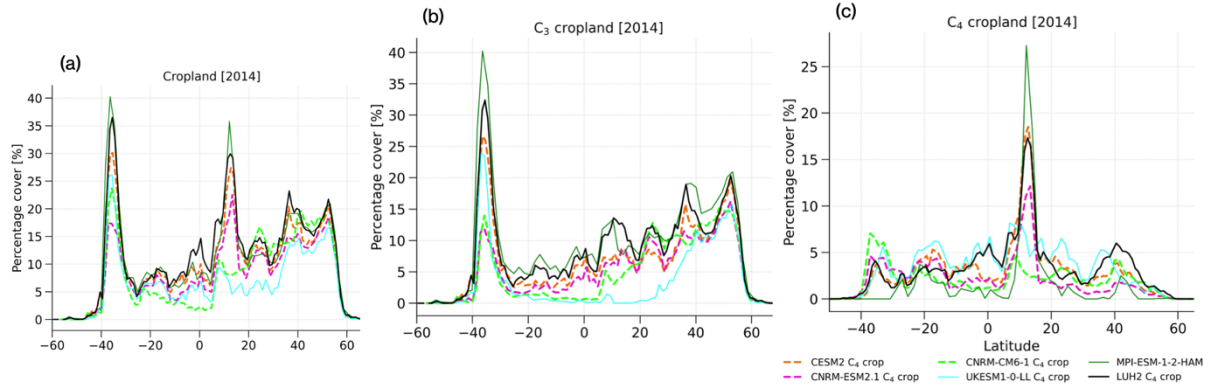


Figure S812— Plot of the spatial variation in cropland area in CMIP6 models and LUH2 for 2014. Total cropland (a), C₃ crop (b), and C₄ cropland (c). The spatial variation of C₃ and C₄ crops obtained by multiplying the total crop cover and the total C₃ and C₄ fraction in CESM2, CNRM-CM6.1 and CNRM-ESM2.1 is consistent with the range of those of the other models and the LUH2 datasets.

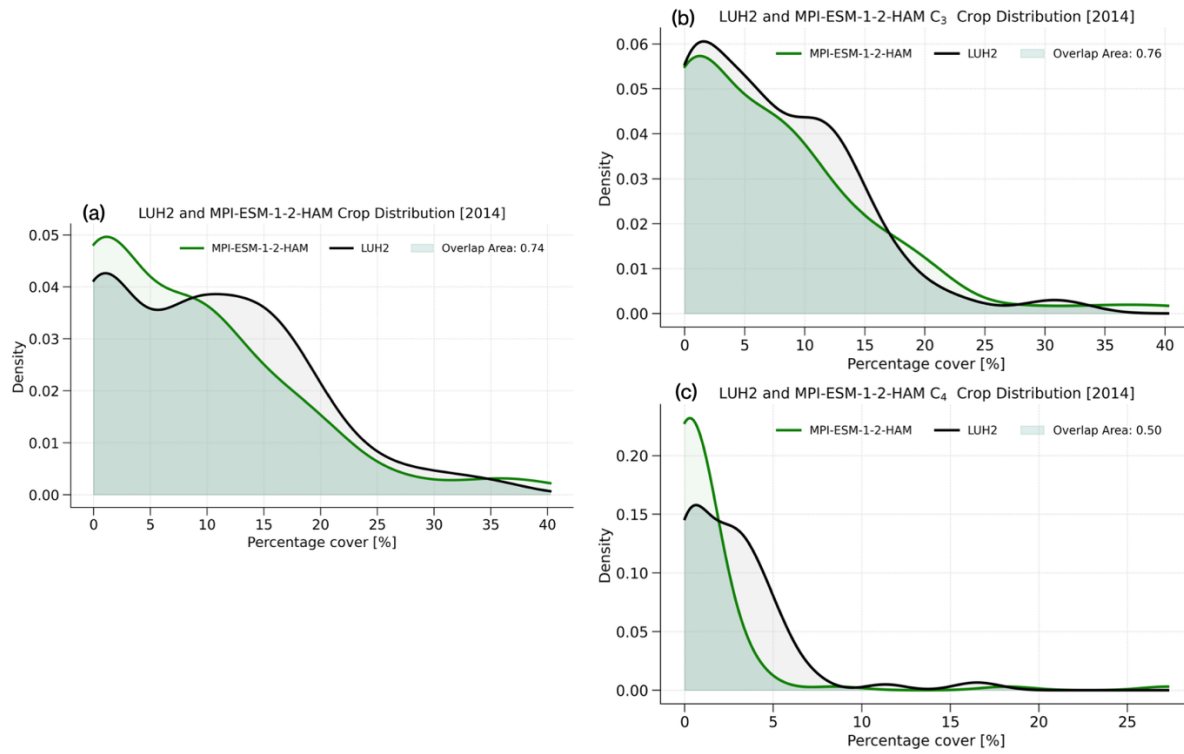


Figure S213 – Kernel density plot of the spatial distribution of crop in and MPI-ESM-1-2-H and LUH2 for 2014. Total cropland (a), C₃ crop (b), and C₄ cropland (c). The overlap area is 74, 76 and 50 % for total, C₃ and C₄ crops between the model and LUH2.

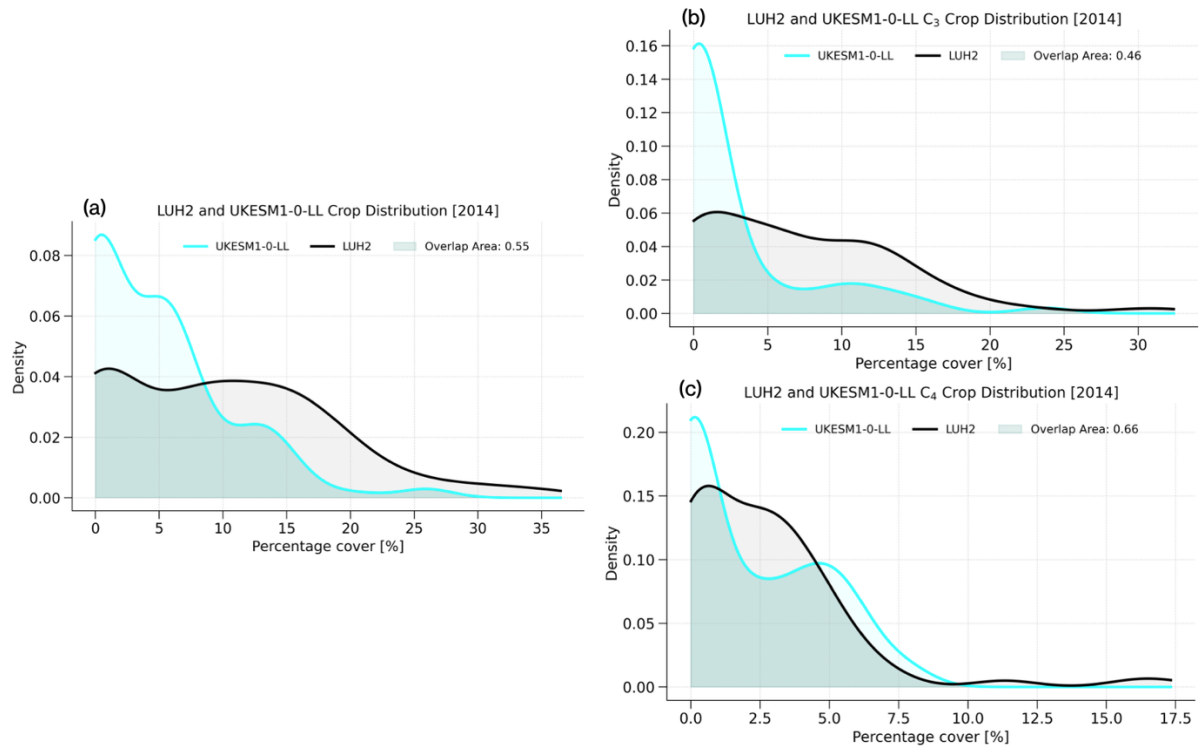


Figure S1014— Kernel density plot of the spatial distribution of crop in and UKESM1 and LUH2 for 2014. Total cropland (a), C₃ crop (b), and C₄ cropland (c). The overlap area is 55, 46 and 66 % for total, C₃ and C₄ crops between the model and LUH2.

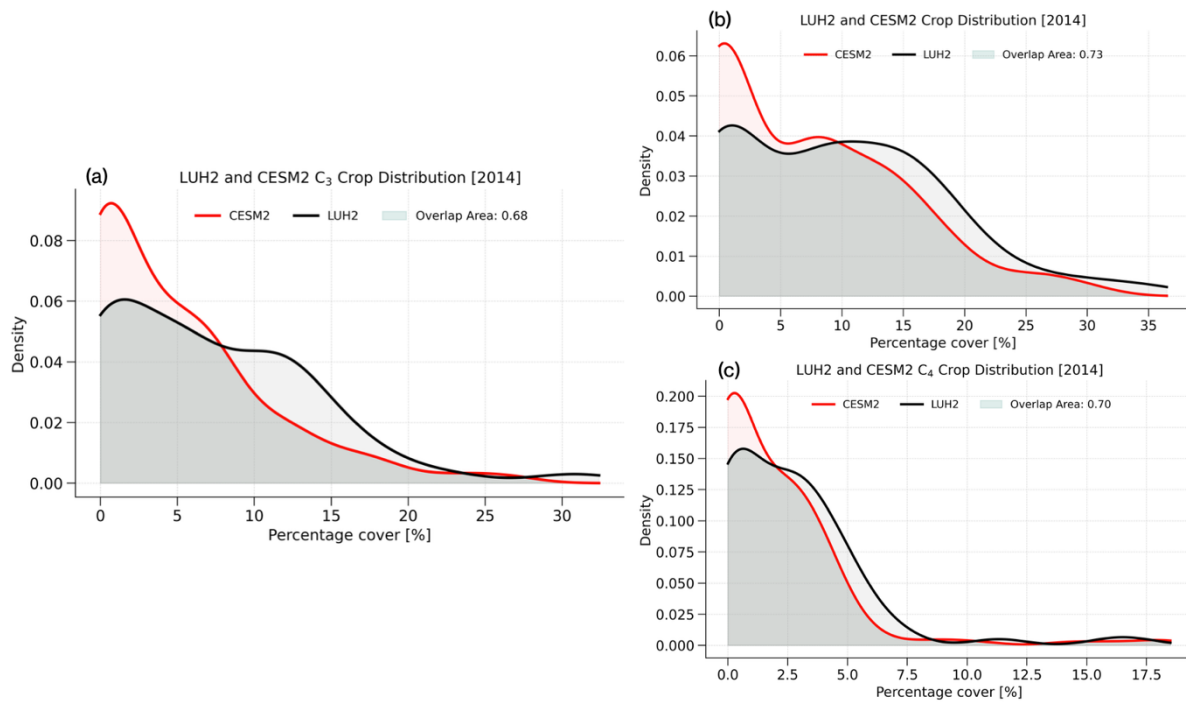


Figure S1145 – Kernel density plot of the spatial distribution of crop in and CESM2 and LUH2 for 2014. Total cropland (a), C₃ crop (b), and C₄ cropland (c). The overlap area is 68, 73 and 70 % for total, C₃ and C₄ crops between the model and LUH2.

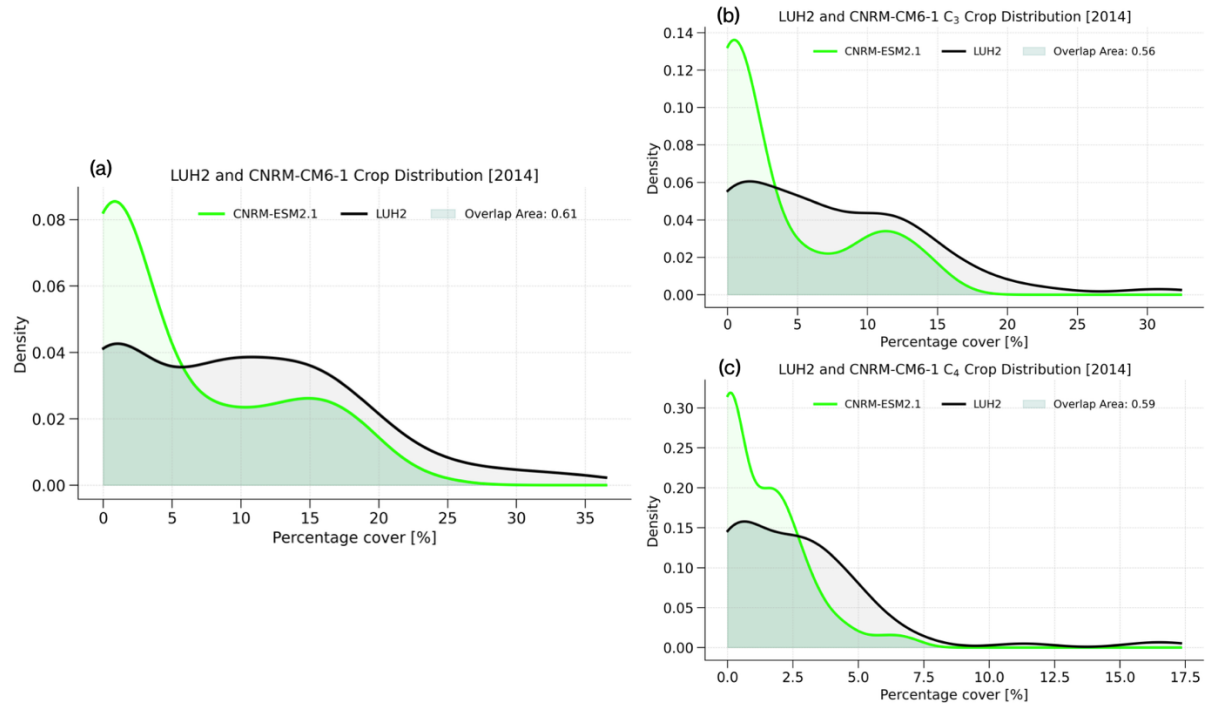


Figure S1246 – Kernel density plot of the spatial distribution of crop in and CNRM-CM6.1 and LUH2 for 2014. Total cropland (a), C₃ crop (b), and C₄ cropland (c). The overlap area is 61, 56 and 59 % for total, C₃ and C₄ crops between the model and LUH2.

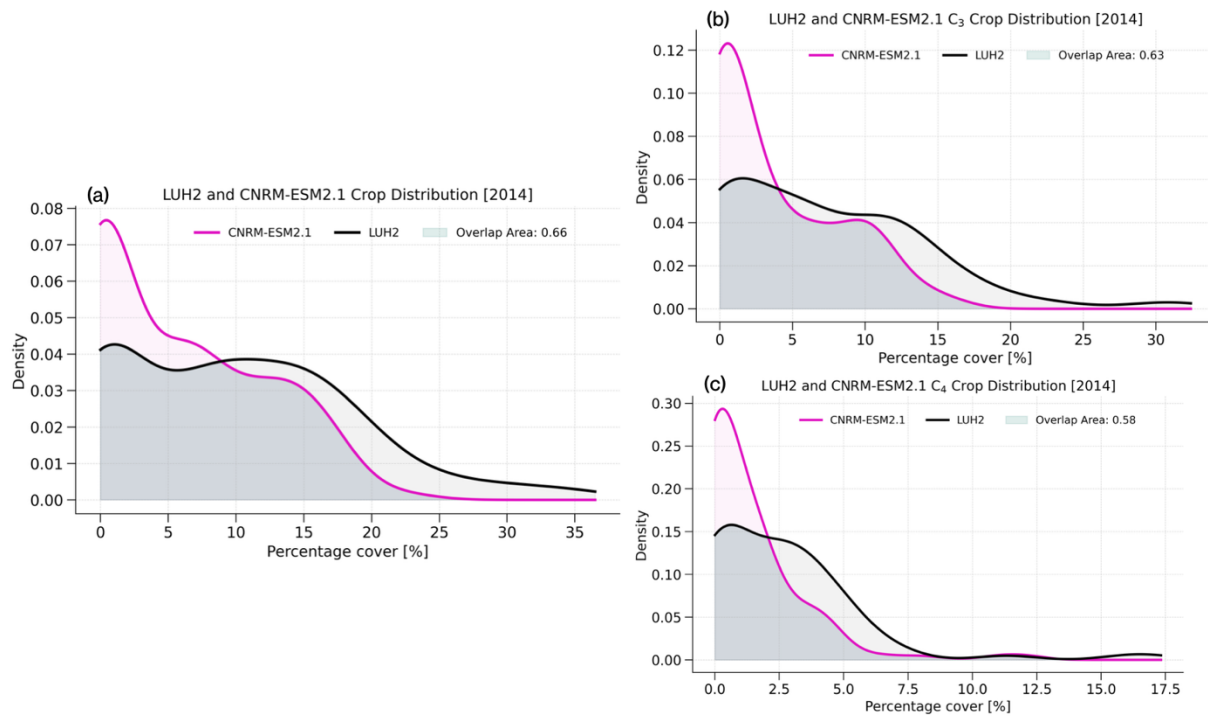


Figure S1347— Kernel density plot of the spatial distribution of crop in and CNRM-ESM2.1 and LUH2 for 2014. Total cropland (a), C₃ crop (b), and C₄ cropland (c). The overlap area is 66, 63 and 58 % for total, C₃ and C₄ crops between the model and LUH2.