



Uncertainties in the land-use flux resulting from land-use change reconstructions and gross land transitions

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Abstract. Land-use and land-cover (LUC) changes are a key uncertainty when attributing changes in measured atmospheric CO₂ concentration to its sinks and sources and must also be much better understood to determine the possibilities for land-based climate change mitigation, especially in the light of human demand on other land-based resources. On the spatial scale typically used in terrestrial ecosystem models (0.5 or 1°) changes in LUC over time periods of a few years or more can include bidirectional changes on the sub-grid level, such as the parallel expansion and abandonment of agricultural land (e.g. in shifting cultivation) or cropland–grassland conversion (and vice versa). These complex changes between classes within a grid cell have often been neglected in previous studies, and only net changes of land between natural vegetation cover, cropland and pastures accounted for, mainly because of a lack of reliable high-resolution historical information on gross land transitions, in combination with technical limitations within the models themselves. In the present study we applied a state-of-the-art dynamic global vegetation model with a detailed representation of croplands and carbon–nitrogen dynamics to quantify the uncertainty in terrestrial ecosystem carbon stocks and fluxes arising from the choice between net and gross representations of LUC. We used three frequently applied global, one recent global and one recent European LUC datasets, two of which resolve gross land transitions, either in Europe or in certain tropical regions. When considering only net changes, land-use-transition uncertainties (expressed as 1 standard deviation around decadal means of four models) in global carbon emissions from LUC (E_{LUC}) are ± 0.19 , ± 0.66 and ± 0.47 Pg C a⁻¹ in the 1980s, 1990s and 2000s, respectively, or between 14 and 39 % of mean E_{LUC} . Carbon stocks at the end of the 20th century vary by ± 11 Pg C for vegetation and ± 37 Pg C for soil C due to the choice of LUC reconstruction, i.e. around 3 % of the respective C pools. Accounting for sub-grid (gross) land conversions significantly increased the effect of LUC on global and European carbon stocks and fluxes, most noticeably enhancing global cumulative E_{LUC} by 33 Pg C (1750–2014) and entailing a significant reduction in carbon stored in vegetation, although the effect on soil C stocks was limited. Simulations demonstrated that assessments of historical carbon stocks and fluxes are highly uncertain due to the choice of LUC reconstruction and that the consideration of different contrasting LUC reconstructions is needed to account for this uncertainty. The analysis of gross, in addition to net, land-use changes showed that the full complexity of gross land-use changes is required in order to accurately predict the magnitude of LUC change emissions. This introduces technical challenges to process-based models and relies on extensive information regarding historical land-use transitions.

1 Introduction

Due to the increasing demand of a growing population for food and fibre, as well as for bioenergy, greater anthropogenic pressures on the global land area are expected. Today, carbon dioxide (CO₂) emissions resulting from land-use and land-cover (LUC) change are the second largest contributor to anthropogenic emissions to the atmosphere after fossil fuel combustion (Le Quéré et al., 2015a), and they are a factor that is associated with large uncertainties. LUC and its changes include the processes when land is converted from one land-cover type to another (e.g. the conversion of forest to cropland or grasslands to pastures), and the effects of LUC related to the management of the land, such as cropping practices, fertiliser use, irrigation and different types of tillage. LUC changes affect the cycling of carbon (C), energy, water and other nutrients (phosphorous, nitrogen), in many cases enhancing greenhouse gas (e.g. CO₂, N₂O, CH₄) emissions from agricultural soils and pastures when compared to natural vegetation and altering species composition. These changes go hand in hand with altered characteristics such as surface albedo, surface aerodynamic roughness and rooting depth (Pongratz et al., 2010).

Conversions from natural vegetation to croplands and pastures generally reduce C stored in vegetation (Baccini et al., 2012), decrease soil C stocks in croplands but not in pastures (Guo and Gifford, 2002; McLauchlan, 2006) and, unless fertilised, reduce soil nitrogen (N) pools (McLauchlan, 2006). The alteration of C and N pools are mainly a result of initial deforestation and of the decreased litter input due to biomass extraction upon harvest and accelerated soil decomposition rates, the latter being stimulated through management practices such as tillage or a changed microclimate at the soil surface. However, in some regions croplands show increased C sequestration potential compared to the natural vegetation owing to enhanced growth under improved agricultural practices including fertilisation and irrigation (Ciais et al., 2010; Schulze et al., 2010). Legacy fluxes can change C due to, e.g., an imbalance between reduced litter input and decomposing dead biomass and affect LUC emissions over decades or more (Gasser and Ciais, 2013; Houghton, 2010; Krause et al., 2016; Pugh et al., 2015). During vegetation recovery on abandoned agricultural land, secondary land ecosystems sequester C due to regrowing vegetation and reaccumulation of C in soils. These LUC-related processes determining regional sources and sinks of C entailed a global total net C flux to the atmosphere over the past centuries (deB Richter and Houghton, 2011; Houghton et al., 2012; McGuire et al., 2001; Le Quéré et al., 2015a).

A number of studies have recently highlighted the importance of different definitions when assessing the net carbon flux from LUC (E_{LUC}) related to the fact that in different studies different LUC component fluxes are included in the overall E_{LUC} calculation (Gasser and Ciais, 2013; Pongratz et al., 2014). Likewise, it is important to consider whether or not historical effects of environmental change are included in assessments of cleared C stocks as part of E_{LUC} . Less focus so far has been put on the explicit datasets of historical land use employed (Le Quéré et al., 2014). A limited number of historic LUC reconstructions are available on a global scale, mostly at 0.5° spatial resolution (Hurtt et al., 2011, 2016; Kaplan et al., 2012; Klein Goldewijk, 2016; Klein Goldewijk et al., 2011; Olofsson and Hickler, 2008; Pongratz et al., 2008; Ramankutty and Foley, 1999), two of which are very similar (datasets of Hurtt et al., 2011 and Klein Goldewijk et al., 2011 are consistent when the corresponding versions are compared). On a continental scale some higher-resolution reconstructions exist, for instance for Europe (Fuchs et al., 2015b; Kaplan et al., 2009; Williams, 2000). Most reconstruction approaches combine information on current and recent historical LUC from national statistics with estimates of global population distribution and growth as the main driver of historical LUC. Model assumptions are made to fill data gaps and extrapolate the available information to create sub-national patterns, and therefore large uncertainties arise both from the original data sources and the modelling assumptions (see, e.g., Klein Goldewijk and Verburg, 2013). However, reconstructions on continental scales are able to use a more data-driven approach (e.g. Fuchs et al., 2013, 2015c) compared to global land reconstructions, since the data availability is often better for these study areas.

Most historical LUC reconstructions focus on the difference in net area under natural, cropland or pasture vegetation cover at a grid location between two time steps (net land changes) instead of explicitly showing the sum of the absolute value of all land transitions occurring on a sub-grid scale (gross land changes). In particular over coarser grid resolutions, gross land-cover changes allow a deeper view of LUC, tracking land conversion events such as the parallel expansion and abandonment of agricultural land, e.g. as in shifting cultivation (cycle of cutting forest for agriculture and abandoning it after some years of usage, followed by a period of fallow with regrowing forests). This entails altered biogeochemical dynamics within different subsections of a grid cell; e.g. secondary land acts as a C sink during vegetation regrowth, while additional land clearing leads to a relatively rapid loss of C stocks in vegetation and soils, along with other changes in vegetation composition, nutrients and biogeophysical properties (e.g. Houghton et al., 2012). Accounting in ecosystem models separately for the effects of individual transitions (e.g. 10 % of an area converted from natural vegetation to cropland while another 10 % of crop-

land is abandoned for regrowth over the same time period) will therefore lead to a very different response of ecosystem states and fluxes compared to the effects of net changes, which in this case would be zero.

The availability of land-use information including gross land transitions is limited due to a lack of reliable historical information to determine it. However, a few datasets exist representing gross land transitions, such as the global dataset by Hurtt et al. (2011) (recently updated; see Hurtt et al., 2016), who provide model estimates of shifting cultivation in certain tropical areas based on a map of Butler (1980) and assumed land rotation rates. Fuchs et al. (2015b) recently estimated gross land changes for Europe over the 20th century based on empirical evidence. As the number of gross land transitions can greatly exceed the number of net transitions at spatial resolutions typically employed for global studies, neglecting these can lead to a serious underestimation of LUC dynamics with implications for biogeochemical, ecological and environmental assessments (Fuchs et al., 2015a, b; Stocker et al., 2014; Wilkenskjeld et al., 2014). Earlier studies revealed significant differences when ecosystem C dynamics were simulated when accounting for gross land changes in areas of shifting cultivation in addition to net changes as specified by Hurtt et al. (2011) (e.g. Shevliakova et al., 2013; Stocker et al., 2014; Wilkenskjeld et al., 2014). Others have implemented their own assumptions on spatial distribution and the rotation scheme under shifting cultivation and combined these with C-cycle calculations (Olofsson and Hickler, 2008; Stocker et al., 2014).

In this study we use a state-of-the-art dynamic global vegetation model (DGVM) to calculate ecosystem C stocks and fluxes in response to different LUC reconstructions (1) to explore the consistency of different LUC representations and to quantify the uncertainty in ecosystem C stocks and fluxes, including E_{LUC} , resulting from the different reconstructions and (2) to quantify the effect of accounting for gross land transitions in addition to net changes in LUC. We use five historical LUC reconstructions, four of which are global and one of which is only for the European domain. One of the global datasets represents gross transitions due to shifting cultivation in certain tropical regions, and the European dataset represents gross transitions from all sources in Europe. We apply a model with a representation of LUC and changes therein, including a number of crop functional types and C–N dynamics in natural vegetation and crops. We exclude wood harvest as a form of forest management that can be represented as gross land transitions from our analysis as, although national data on wood harvest are available, its parameterisation in models is poorly constrained on a global scale; e.g. the effects strongly depend on assumptions on the harvest type (clear cut, selective logging, or a mixture of both) or assumptions regarding turnover times of harvested C (Wilkenskjeld et al., 2014).

2 Methods

2.1 Land-use datasets

For the global scale, three historical LUC datasets were selected that are frequently used for ecosystem modelling studies, as well as one recently released dataset that is expected to be frequently used in the future. These datasets run at least from 1700 to present and are also the basis for future LUC scenarios (e.g. van Asselen and Verburg, 2013; Hurtt et al., 2011). For Europe we additionally considered one recently published dataset running from 1900 to 2010. Table 1 provides an overview of the LUC datasets and their characteristics.

Ramankutty and Foley (1999) (RAMA) published changes in cropland area for the period 1700 to 1992 at a 5 min global resolution. The dataset was built based on historical cropland inventory data at national and subnational levels in combination with a remote-sensing-derived cropland map for 1992. The algorithm to hindcast LUC distributed the historical cropland area within political units, i.e. the 1992 cropland areas are scaled for each political unit so that the cropland total matches historical inventory data. Therefore, the reconstructed changes in historical croplands are consistent with the history of human settlement and patterns of economic developments, although they do not resolve changes in LUC dynamics below the smallest spatial entity in the inventory data. The analysis was revised in 2012 so that it also accounts for pasture areas, and the dataset was extended until 2007 at a $0.5^\circ \times 0.5^\circ$ spatial resolution (Ramankutty, 2012). Natural areas are calculated as the remainder of cropland and pasture areas. Here we apply the revised $0.5^\circ \times 0.5^\circ$ version.

The History Database of the Global Environment (HYDE) 3.1.1 (Klein Goldewijk et al., 2010, 2011) provides spatially gridded maps of cropland and pastures at a 5 min resolution for the period 10 000 BC to AD 2000. Here, historical population data and national and subnational statistics of change in agricultural area (mainly the United Nations Food and Agriculture Organization, FAO, supplemented with numerous other statistics) were combined to a land-use per capita relationship. Land use was then allocated for the present day based on satellite-derived land cover for 2000 and for the past by a combination of this base map with a number of weighting and suitability factors such as population density and habitat information on soil suitability, distance to rivers, slopes, etc. The temporal resolution is 10 years for the historical period after 1700 and annual after 2000. The HYDE dataset used here was extended until 2005 (Klein Goldewijk et al., 2015) and later until 2013 in the 2014 global carbon budget analysis (Le Quéré et al., 2015a).

The Hurtt et al. (2011) Land Use Harmonization v1 (LUH1) database is based on the land-use data of HYDE (Klein Goldewijk et al., 2010, 2011) for the historical period (1500–2005) and combines these with national statistics of historical wood harvest and assumptions regarding shift-

Table 1. Raw characteristics of LUC datasets used in this study^a.

Land-use model		Time period			Time steps	Representation of LUC transitions	Spatial coverage	Original spatial resolution	Land-cover classes
	Model version and reference	From	to						
LUH1	1500–2005: LUH1 (Hurtt et al., 2011); extension until 2014: Le Quéré et al. (2015b) ^b	AD 1500	2014 ^c	annual	net/gross	global	0.5°	cropland, pasture, primary natural vegetation, secondary natural vegetation, urban	
RAMA	1700–1992: Ramankutty and Foley (1999); extension until 2007: Ramankutty (2012)	AD 1700	2007	annual	net	global	0.5°	cropland, pasture, primary natural vegetation, secondary natural vegetation, urban	
HYDE	10 000 BC to AD 2000: HYDE3.1.1 (Klein Goldewijk et al., 2010, 2011); extension until 2005: see Klein Goldewijk et al. (2015); extension until 2013: see Le Quéré et al. (2015a)	10 000 BC	2013	annual ^d	net	global	5 arcmin	cropland, pasture, natural vegetation ^e	
LUH2	850–2015: LUH2 v2, release 14 Oct 16 (Hurtt et al., 2016)	AD 850	2015	annual	net ^f	global	0.25°	C3/C4 annual/perennial crops, C3 N-fixing crops, managed pasture, rangeland, primary natural vegetation, secondary natural vegetation, forest, non-forest, urban	
HILDA	HILDA v2.0 (Fuchs et al., 2015b)	AD 1900	2010	decadal	net ^g /gross	EU27 ^h plus Switzerland	1 km	cropland, grassland (incl. managed pastures and shrubland), forest, settlements, water, other land (glaciers, sparsely vegetated areas, beaches and water bodies)	

^a Note that these datasets are preprocessed into more consistent characteristic sets before carrying out simulations – see the “Methods” section and Table 2. ^b Note that this version of LUH1 is based on an early version of HYDE 3.2, which is different from HYDE version 3.1.1 as used below; see the “Methods” section. ^c End date of historical land-use dataset. ^d Decadal data until 2000. ^e Natural vegetation is calculated as a remainder. ^f Only the net dataset was used in this study (see the “Methods” section). ^g The HILDA net dataset used in this study was derived from the gross dataset (see the “Methods” section). ^h European Union 2007–2013.

ing cultivation in certain tropical regions. Additional assumptions were made regarding the prioritisation of land for conversion and logging, the wood harvest spatial patterns and the residence time of land in agricultural use in shifting cultivation areas. LUH1 data provide fractional data on cropland, pasture, primary and secondary vegetation as well as gross transitions between land-use states based on shifting cultivation at a $0.5^\circ \times 0.5^\circ$ spatial resolution. Secondary land is defined as natural land that was previously used for agriculture and is recovering from this disturbance. Shifting cultivation is implemented as bidirectional LUC change with an assumed rotation period of 15 years, corresponding to an annual turnover rate of 6.7% of the area (Hurtt et al., 2011). Although the history of shifting cultivation is not known, it is today present mainly in tropical regions; therefore, in the LUH1 dataset it is limited to certain tropical regions for the historical period (Fig. S1 in the Supplement) Olofsson and Hickler, 2008). The LUH1 dataset was extended until 2014 for the 2015 global carbon budget analysis (Le Quéré et al., 2015b), using an early version of HYDE 3.2 as the basis (now published in final version as Klein Goldewijk, 2016) and following the same methodology as described in Hurtt et al. (2011). The version of LUH1 used in this study is there-

fore a more recent development than that used for CMIP5 experiments (Taylor et al., 2012), but an earlier version than the very recent LUH2 release (Hurtt et al., 2016). As LUH1 is a modelled product that is based on the underlying HYDE database, these products are very similar when the corresponding versions of each dataset are considered (Hurtt et al., 2011). Note that the version of HYDE used for our study (version 3.1.1, see above) is not the same as the version of HYDE that underlies the LUH1 data used here (early version of HYDE 3.2). HYDE 3.1.1 and 3.2 differ in terms of driving population data and the algorithms used (Klein Goldewijk et al., 2016). The HYDE and LUH1 data used in this study therefore differ in both their spatial and temporal distribution of land-use fractions (Fig. 1; Klein Goldewijk et al., 2016).

In addition, we used the very recent release of the Land Use Harmonization v2 (LUH2; Hurtt et al., 2016) that was developed for CMIP6 intercomparison project (Eyring et al., 2016). This global dataset covers the period 850–2015 and follows the same methodology as LUH1 described above, but uses HYDE version 3.2 (Klein Goldewijk, 2016) as a basis, along with updated attributes on wood harvest and shifting cultivation, a higher spatial resolution of 0.25° , more de-

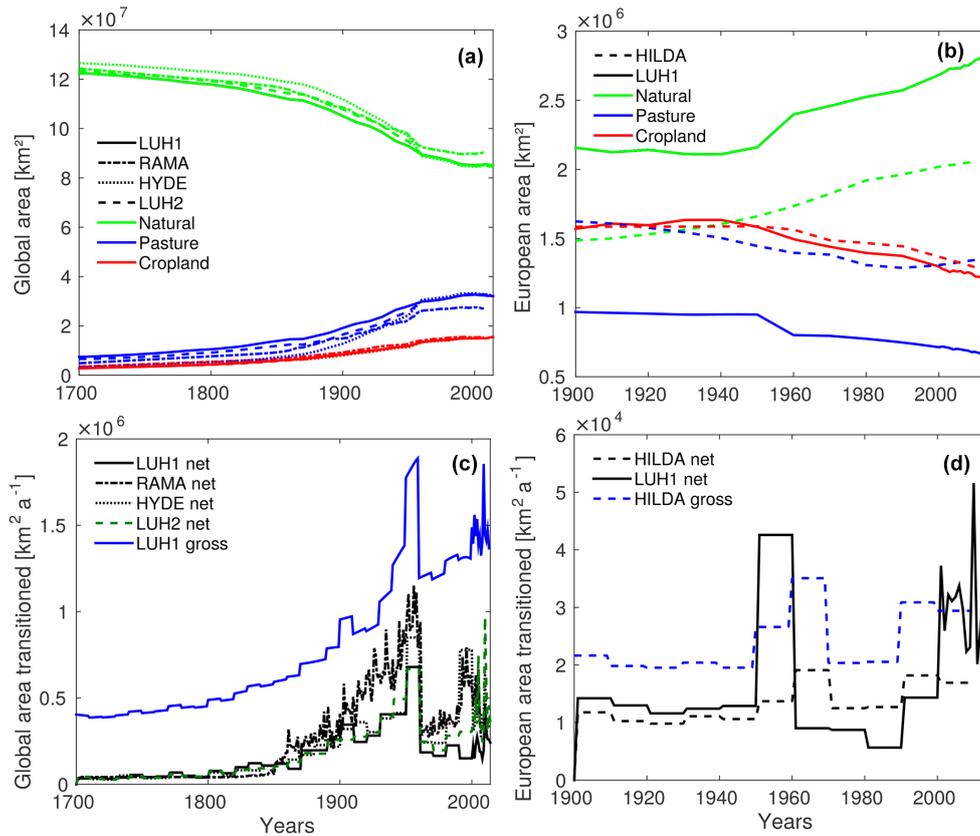


Figure 1. Land-use types and transitions in global and European (EU27 + CH) LUC reconstructions. The evolution of absolute land area of croplands, pastures and natural vegetation (including barren land) in different (a) global historical land-use reconstructions (LUH1: solid line; RAMA: dash-dotted line; HYDE: dotted line; LUH2: dashed line) and (b) European land-use reconstructions (HILDA: dashed line; LUH1: solid line). Land area experiencing gross and net land transitions on a global scale (c) and for Europe (d). Note the change to annual resolution in the LUH1 reconstruction after the year 2000.

tailed land-use transitions and additional land management information (such as irrigation and fertiliser use). From the LUH2 dataset, only LUC states and net land transitions aggregated to a spatial resolution of $0.5^\circ \times 0.5^\circ$ were considered in this study, as a very new dataset that very likely will frequently be applied for modelling studies in the future. As LUC patterns of LUH2 are directly based on HYDE3.2, these two datasets are very similar in their land-use information (data not shown), and therefore resulting ecosystem C stocks and fluxes are expected to be very similar. The detailed LUC gross transitions provided by LUH2 could not be preprocessed in time to be used here.

The only non-global LUC dataset considered here, the HIstoric Land Dynamics Assessment (HILDA; Fuchs et al., 2013, 2015b), reconstructs gross and net land changes for the EU27 (EU from 2007 to 2013) plus Switzerland at a 1 km spatial resolution (Fig. S1). Net land conversions are based on national statistics. To account for gross changes, empirical evidence (mostly time series of large-area LUC maps and national surveys) on historic gross LUC changes was aggregated to derive an overall gross / net ratio per LUC class and

a relative weighted land conversion matrix, both of which were applied to national net change data. The allocation of LUC fractions to grid cells was done based on probability maps for each LUC class, forest volume stock maps and large-scale historic LUC maps (Fuchs et al., 2015c). An aggregated version of the CORINE2000 land-cover dataset was used as base map for the year 2000. The initial LUC dataset that was built based on empirical evidence covers 1950–2010 in decadal steps but was extrapolated back to 1900 to assess the long-term impacts of changes in LUC. For each time step the 1 km grid cells were classified as being dominated by settlement, cropland, forest, grassland (including managed pastures), other land (glaciers, sparsely vegetated areas, beaches and water bodies) or water. Here, we consider only the gross LUC reconstruction of HILDA (Fuchs et al., 2015b) and derive net LUC changes from gross land transitions. In the original HILDA net LUC reconstruction (Fuchs et al., 2015b), the results differ spatially from our net reconstruction due to the use of different land allocation mechanisms under net and gross changes in their analysis.

2.2 LPJ-GUESS ecosystem model

We use the LPJ-GUESS (Lund–Potsdam–Jena General Ecosystem Simulator) DGVM (Sitch et al., 2003; Smith et al., 2001) with updates for land-use change (Lindeskog et al., 2013) and C–N coupling in natural vegetation and crops (Olin et al., 2015; Smith et al., 2014) allowing for the simulation of nitrogen limitation on plant and crop development. Three distinct land-use types are used (natural vegetation, pasture and cropland) with natural vegetation modelled by nine woody and two grass plant functional types (PFTs) (as in Smith et al., 2014), which are distinguished in terms of their bioclimatic preferences, photosynthetic pathways and growth strategies. Vegetation structure, dynamics and competition between age cohorts of a PFT population are explicitly represented in the LPJ-GUESS model. Croplands are represented by three crop functional types (CFTs) that are parameterised using information on summer wheat, winter wheat and maize, with crop-specific processes including dedicated carbon allocation and phenology; explicit sowing and harvest representation; irrigation; fertilisation; and cover crops represented by grass growing in between cropping periods (Olin et al., 2015). Pastures are modelled using competing C3 and C4 grass PFTs, where each year 50 % of the C and 12.5 % of the N in above-ground biomass was removed as a representation of grazing (Krause et al., 2016; Lindeskog et al., 2013).

In LPJ-GUESS, upon conversion of natural land to cropland and pastures, the natural vegetation is cleared and 97 % of wood (of which stem wood is 65 % and branches and coarse roots are 32 %), and 71 % of leaf biomass is harvested. Out of the harvested stem wood, one-third goes to a product pool with a turnover time of 25 years. The rest of the harvested biomass is oxidised and released to the atmosphere, while the remaining biomass enters the litter pool (see Lindeskog et al., 2013). In reductions of the natural vegetation area, young stands (but older than 15 years, the rotation period in shifting cultivation assumed by Hurtt et al., 2011) are converted before older stands. Following agricultural abandonment, natural vegetation recolonises the land in a typical succession from herbaceous to woody plants if environmental conditions are suitable for tree growth. Competition for resources and light among age cohorts of woody PFTs is simulated directly through gap dynamics (see, e.g., Bugmann, 2001).

The model has been evaluated extensively and has demonstrated skills in capturing large-scale vegetation patterns (Hickler et al., 2006, 2012) and dynamics of the terrestrial carbon cycle (Ahlström et al., 2012; Morales et al., 2005; Olin et al., 2015; Piao et al., 2013; Pugh et al., 2015; Smith et al., 2014). The carbon flux response was close to the ensemble mean in a recent intercomparison of nine dynamic global vegetation models (Sitch et al., 2015).

2.3 Simulation protocol

LPJ-GUESS was run at a $0.5^\circ \times 0.5^\circ$ resolution with simulations beginning in the year 1700. CRU TS 3.21 historical global climate data (University of East Anglia Climatic Research Unit, CRU, 2013) was used for the period 1901–2014. Climate data for 1700–1900 were provided by repeating 1901–1930 CRU climate with de-trended temperature data. Climate data for 2014 were repeated for the year 2015. Atmospheric CO₂ forcing was provided from observations from ice cores and, later in the 20th century, atmospheric measurements (Tans and Keeling, 2015), with a value of 286.4 ppmv used from 1700 until the beginning of this dataset in 1860 and a final atmospheric mixing ratio of 399.0 ppmv in 2015. Simulations were spun-up for 500 years using land-use fractions and the CO₂ mixing ratio from the first simulation year and de-trended climate data of the first 30 simulation years, with a longer spin-up for soil carbon pools (see Smith et al., 2014). Model spin-up was therefore identical for net and gross land changes. In order to assign cropland areas to crop functional types, global crop cover fraction was partitioned based on Portmann et al. (2010) and mapped to LPJ-GUESS crop types, as described in Olin et al. (2015). Crop type fractions were held constant throughout the simulations. Where cropland was expanded into a hitherto uncropped grid cell, average CFT fractions of the nearest neighbouring cropland cells were used to populate it. Past values of global nitrogen deposition were taken from simulations from Lamarque et al. (2010, 2011) and nitrogen fertilisation of crops was estimated as in Zaehle et al. (2011). LPJ-GUESS simulations are summarised in Table 2.

Global simulations starting in 1700 were carried out using the four net and one gross LUC dataset (LUH1 net, RAMA net, HYDE net, LUH2 net, LUH1 gross), and those for Europe starting in 1900 used two net and one gross LUC dataset (HILDA net, LUH1 net, HILDA gross). For these, all LUC input data were aggregated to a 0.5° spatial resolution and decadal HILDA, HYDE and LUH1 LUC data were interpolated linearly to annual time steps. LPJ-GUESS uses annual land-use states of the classes “cropland”, “pasture”, “natural vegetation” and “barren land” (no vegetation; e.g. water- or ice-covered) as input for net LUC runs, which are complemented for gross LUC runs by annual gross transitions for each combination of two land-use classes. Land-use states of RAMA, HYDE and LUH2 were used directly. To generate net transitions from LUH1, annual land-use states were derived from land-use states in 1700 and gross transitions from 1700 to 2014. HILDA land-use matrices providing land-use states and transitions together in the form of an integer land-use category were translated to annual land-use states and gross transitions for each combination of two land-use classes. Although some of these LUC products represent changes between forested and non-forested land, in the simulations done here, only the changes between the classes croplands, pastures (FAO category “permanent pasture”), natural

Table 2. Overview of LPJ-GUESS simulations carried out as part of this study.

Abbreviation	First year	Last year	Representation of LUC transitions	Spatial coverage
LUH1	1700	2014	gross	global
	1700	2014	net	global
	1700	2014	LUC fixed to 1700	global
RAMA	1700	2007	net	global
	1700	2007	LUC fixed to 1700	global
HYDE	1700	2013	net	global
	1700	2013	LUC fixed to 1700	global
LUH2	1700	2015	net	global
	1700	2015	LUC fixed to 1700	global
HILDA	1900	2010	gross	EU27 + CH
	1900	2010	net	EU27 + CH
	1900	2010	LUC fixed to 1900	EU27 + CH
LUH1	1900	2014	net	EU27 + CH
	1900	2014	LUC fixed to 1900	EU27 + CH

vegetation and barren land (available for LUH1, LUH2 and HILDA) were considered; the composition of natural vegetation was simulated directly by LPJ-GUESS. Primary and secondary vegetation as in LUH1 and LUH2 (wood harvest) and the forest class in HILDA and LUH2 were represented by natural vegetation. The HILDA LUC classes of settlements, water and other land were aggregated to the LUC class barren. The grassland class in HILDA comprises both meadows and natural grasslands, which were both represented using the pasture class in LPJ-GUESS, a reasonable assumption for Europe due to the small area of truly natural, unmanaged grassland in Europe (Wilkins et al., 2003). For global simulations a land mask was used that includes only cells of the ice-free land area for which all four global LUC datasets provide data (58 790 cells). For Europe, all 0.5° grids that contained at least one HILDA cell were used (2486 cells).

We examine differences caused by the different LUC reconstructions on net land-use flux (E_{LUC}), deforestation flux, net primary productivity (NPP) and ecosystem C stocks in vegetation and soils. E_{LUC} is calculated as the difference between a model simulation with transient historical LUC change (gross or net LUC change) and a simulation with constant LUC distribution as in the first simulation year (Table 2, LUC fixed to 1700/1900). All simulations are driven by varying climate, atmospheric CO_2 mixing ratio, N deposition and N fertilisation (see, e.g., Le Quéré et al., 2015; Pongratz et al., 2014). This method includes the effects of LUC and changes therein interacting with climate and atmospheric CO_2 and is consistent with definition 1 of Gasser and Ciais (2013) and D3 of Pongratz et al. (2014). In the same way, the effect of accounting for LUC on NPP and C stocks in vegetation and soils was calculated as the difference between the simulation with gross or net LUC changes and the respective reference simulation. Soil C includes both C in soils and litter. The deforestation flux is the C released upon land conversion only. In the calculation of net

cumulative E_{LUC} for global simulations the first 50 simulation years were ignored because of high carbon fluxes in the first decades of gross simulations, which reflected a re-equilibration under LUC including gross transition rates (i.e. shifting cultivation) that were not part of model spin-up and effectively reflect emissions from shifting cultivation that occurred before 1700 (see, e.g., Stocker et al., 2014). Because gross transitions in Europe do not follow a systematic rotation such as shifting cultivation areas in the global simulations, this effect is not so directly applicable or apparent here and cumulative E_{LUC} was calculated starting in 1900. We restrict our analysis of the effects of using different LUC reconstructions to the global scale, which has direct relevance for the global carbon budget; a detailed analysis of regional differences and processes is beyond the scope of this study.

3 Results

3.1 Historical land transitions

The most pronounced change in global vegetation cover over the historical period is the deforestation of natural areas for conversion into croplands and pastures (Fig. 1a), progressing at fairly low rates during the first decades after 1700, followed by a steadily increasing trend from about 1860 until a slow-down and stabilisation sets in after about 1960. Total land area transitions before 1850 (Fig. 1c) are below $100\,000\text{ km}^2\text{ a}^{-1}$, from which they steadily increase with a rate of an additional $ca. 6100 \pm 1100\text{ km}^2\text{ a}^{-1}$ under transition each year (average of four LUC datasets 1850–1960, Table 3). Transitions in all four LUC reconstructions peak between 1950 and 1960 when deforestation due to expansion of agriculture in the tropics and pasture expansion in grass- and shrub-dominated biomes was highest in the LUC reconstructions (Fig. S2). After the 1960s all four LUC reconstructions assume continued deforestation in the tropics

Table 3. Changes in C stocks and fluxes in four global reconstructions of net and gross LUC changes. Land-use change flux (E_{LUC}) and cumulative land-use flux E_{LUC} ; net primary productivity (NPP); C stocks in vegetation and soils. Values are always given as 10-year averages (except LUC areas and cumulative flux).

	Averaging period	Calculation	Unit	LUH1 net	RAMA net	HYDE net	LUH2 net	Average and uncertainty for four net LUC models	LUH1 gross	difference LUH1 (gross-net)
Natural area	1700	A_{nat}	10^6 km^2	122.48	124.35	126.54	123.23	124.15 ± 1.77	122.48	0
Natural area	2007	A_{nat}	10^6 km^2	85.39	90.31	84.97	85.01	86.42 ± 2.60	85.39	0
Pasture area	1700	A_{pas}	10^6 km^2	7.45	4.89	3.32	6.61	5.57 ± 1.84	7.45	0
Pasture area	2007	A_{pas}	10^6 km^2	32.38	27.04	32.81	32.73	31.24 ± 2.81	32.38	0
Cropland area	1700	A_{crop}	10^6 km^2	2.80	3.49	2.86	2.89	3.01 ± 0.32	2.80	0
Cropland area	2007	A_{crop}	10^6 km^2	14.96	15.38	14.95	14.99	15.07 ± 0.21	14.96	0
Change in natural area	1700–2007	A_{nat}	10^6 km^2	-37.09	-34.04	-41.57	-38.22	-37.73 ± 3.11	-37.09	0
Change in pasture area	1700–2007	A_{pas}	10^6 km^2	+24.93	+22.15	+29.49	+26.12	$+25.67 \pm 3.04$	+24.93	0
Change in cropland area	1700–2007	A_{crop}	10^6 km^2	+12.16	+11.89	+12.09	+12.10	12.06 ± 0.12	+12.16	0
Total area under transition	1700–2007	ΣA_{trans}	10^6 km^2	51.92	78.59	64.62	57.28	63.10 ± 11.56	243.80	+191.88
Change in area under transition	1850–1960	δA_{trans}	$\text{km}^2 \text{ a}^{-1}$	+5137	+7259	+6703	+5177	6069 ± 1077	+5549	+412
E_{LUC}	1750–2007	$E_{LUC,Net}/Gross$	Pg C a^{-1}	0.81	0.87	0.89	0.77	0.84 ± 0.05	0.94	+0.13
	1750–2014	$E_{LUC,Net}/Gross$	Pg C a^{-1}	0.84	–	–	0.80	–	0.96	+0.12
	1980–1989	$E_{LUC,Net}/Gross$	Pg C a^{-1}	1.10	1.40	1.55	1.31	1.34 ± 0.19	1.28	+0.18
	1990–1999	$E_{LUC,Net}/Gross$	Pg C a^{-1}	1.18	1.57	2.65	1.36	1.69 ± 0.66	1.41	+0.23
	1998–2007	$E_{LUC,Net}/Gross$	Pg C a^{-1}	1.17	2.00	2.06	1.26	1.62 ± 0.47	1.38	+0.20
	2005–2014	$E_{LUC,Net}/Gross$	Pg C a^{-1}	1.50	–	–	1.67	–	1.64	+0.14
Cumulative E_{LUC} from 1750	2007	$\Sigma E_{LUC,Net}/Gross$	Pg C	210.02	225.18	228.95	199.12	215.82 ± 13.82	242.04	+32.02
	2014	$\Sigma E_{LUC,Net}/Gross$	Pg C	222.29	–	–	212.70	–	255.27	+32.98
NPP	1700–1709	$NPP_{Net}/Gross$	Pg C a^{-1}	50.18	52.10	52.21	50.43	51.23 ± 1.07	50.04	-0.14
	1998–2007	$NPP_{Net}/Gross$	Pg C a^{-1}	58.90	59.79	60.85	59.14	59.67 ± 0.87	57.49	-1.42
	2005–2014	$NPP_{Net}/Gross$	Pg C a^{-1}	59.95	–	–	60.18	–	58.46	-1.49
Change in NPP due to LUC	1998–2007	$NPP_{Net}/Gross - NPP_{Ref}$	Pg C a^{-1}	-1.92	-3.32	-2.44	-1.96	-2.41 ± 0.65	-3.33	-1.42
	2005–2014	$NPP_{Net}/Gross - NPP_{Ref}$	Pg C a^{-1}	-2.26	–	–	-2.30	–	-3.75	-1.49
Vegetation C	1700–2014	$VegC_{Net}/Gross$	Pg C	435	–	–	444	–	419	-16
	1700–1709	$VegC_{Net}/Gross$	Pg C	464.18	495.70	497.19	467.59	481.17 ± 17.71	461.35	-2.83
	1998–2007	$VegC_{Net}/Gross$	Pg C	414.62	439.10	435.22	424.53	428.37 ± 11.04	380.43	-34.20
	2005–2014	$VegC_{Net}/Gross$	Pg C	421.48	–	–	431.41	–	386.64	-34.84
Change in vegetation C due to LUC	1960–1969	$VegC_{Net}/Gross - VegC_{Ref}$	Pg C	-109.49	-114.12	-108.21	-98.56	-107.60 ± 6.54	-137.06	-27.58
	1998–2007	$VegC_{Net}/Gross - VegC_{Ref}$	Pg C	-140.36	-153.41	-159.42	-134.36	-146.89 ± 11.54	-174.56	-34.20
	2005–2014	$VegC_{Net}/Gross - VegC_{Ref}$	Pg C	-148.43	–	–	-142.58	–	-183.27	-34.27
Soil C	1700–2014	$SoilC_{Net}/Gross$	Pg C	1425.85	–	–	1438.39	–	1420.81	-5.04
	1700–1709	$SoilC_{Net}/Gross$	Pg C	1445.96	1511.65	1516.06	1453.68	1481.84 ± 37.15	1445.58	-0.38
	1998–2007	$SoilC_{Net}/Gross$	Pg C	1404.20	1471.83	1478.08	1419.90	1443.50 ± 36.97	1393.43	-10.77
	2005–2014	$SoilC_{Net}/Gross$	Pg C	1406.78	–	–	1421.70	–	1395.56	-11.22
Change in soil C due to LUC	1998–2007	$SoilC_{Net}/Gross - SoilC_{Ref}$	Pg C	-75.80	-75.04	-73.54	-67.95	-73.08 ± 3.55	-86.57	-10.77
	2005–2014	$SoilC_{Net}/Gross - SoilC_{Ref}$	Pg C	-77.74	–	–	-70.59	–	-88.96	-11.22

at a lower rate and reforestation in Europe and some parts of Northern America following the abandonment of agricultural areas. Transitions around 1960 are believed to be influenced by the LUC reconstruction process, when model assumptions for the historical period before 1960 are merged with the records of the Food and Agriculture (FAO) records available thereafter.

The four global net LUC datasets applied here differ primarily in the total area of pasture and natural vegetation and in the regions in which these are located; RAMA and LUH2, in terms of global absolute area under natural LUC, lie in between LUH1 (lowest area under natural vegetation before 1960s) and HYDE (highest area under natural vegetation before 1960s, see Fig. 1a). Spatial patterns are gen-

erally more similar between LUH1, HYDE and LUH2 than with RAMA. However, the lower amount of natural areas and higher amount of pastures of LUH1 in the southern parts of Russia compared to the other three reconstructions is noteworthy. Other major differences between all four reconstructions occur in eastern Africa, eastern Europe and parts of the US and Canada (maps not shown). After 1960, LUH1, HYDE and LUH2 are very similar (showing major differences only in Australia). While the deforestation trend is shown by all four global LUC reconstructions, the absolute loss rates of natural vegetation differ, with HYDE being 14 % above the average of the other three LUC models (Table 3). For the present day, differences are largest in natural areas and pastures, with RAMA reporting about 6 % more natural

areas and about 18 % less pasture areas in 2007 compared to the other three reconstructions (Fig. 1a; Table 3). Differences in pasture areas occur worldwide, but are somewhat higher in Saudi Arabia, western China, Mongolia and Australia (maps not shown). Before 1950, differences in natural and pasture area between LUH1, LUH2 and HYDE on the one hand and RAMA on the other exist predominantly in eastern Europe, southern parts of Russia and eastern Africa.

In Europe the two historical LUC reconstructions show the expansion of areas with regrowing natural vegetation after 1900 following land abandonment as a result of intensification on high-production cropland (Fig. 1b). Net gain in natural regrowth area from 1900 to 2010 is about 60 000 km² (average of two LUC datasets, Table 4). The rates of total land conversion in Europe over the first half of the 20th century (Fig. 1d) remain at a fairly constant level, with between 10 000 and 15 000 km² being converted each year. Rates of land conversion are only higher between 1950 and 1970 and after 1990.

The European land-use reconstruction HILDA shows the same trend in LUC over the historical period when compared with the same domain extracted from the global LUH1 product (Fig. 1b), but the two datasets disagree notably with respect to the absolute area of natural vegetation and pastures (Table 4). HILDA shows substantially larger fractions of pasture than LUH1 especially in Scandinavia and southern Europe, while LUH1 shows higher pasture fractions than HILDA in central Europe and the Baltic area. The higher areas of pasture in HILDA may result from the fact that natural grasslands are included in the pasture class in HILDA but not in the pasture class of LUH1 (see the “Methods” section). The peak in total land conversion rates around the middle of the century is shown in HILDA in two steps, with slightly higher rates in the 1950s and a maximum in the 1960s (13 700 and 19 100 km² per year) and in LUH1 in one step with a more than doubled rate in the 1950s compared to the previous decades (on average, about 42 600 km² per year). From 1950 to 1960, the LUH1 dataset shows a rapid decrease in pasture area of 1.5×10^5 km² that is mainly reflected in a significant gain in natural areas.

3.2 Effects of different net LUC changes on carbon pools and fluxes

In LPJ-GUESS simulations all four global net LUC reconstructions resulted in similar projections of the land-use change flux E_{LUC} as a source of C on the global scale, with the rate of emission accelerating from the early 1800s (Fig. 2a). Reflecting the time series of the land transitions (Fig. 1c), E_{LUC} peaked in the 1950s with emissions of about 2.0 to 2.6 Pg C a⁻¹. Mean E_{LUC} was 1.2, 2.0, 2.1 and 1.3 Pg C a⁻¹ for LUH1, RAMA, HYDE and LUH2, respectively, at the end of the historical period (1998–2007, Table 3), and cumulative E_{LUC} since 1750 was between 199 Pg C for LUH2 and 229 Pg C for HYDE in 2007 (Fig. 2b;

Table 3). From the four reconstructions, projections based on HYDE (with LUH2 being very close) resulted in the lowest emissions until the early 1900s, probably because of the lowest conversion of natural areas to pastures until this period compared to the other reconstructions (Fig. 1a). Also, when using the HYDE product, a second peak of E_{LUC} of around 2.7 Pg C a⁻¹ occurred in the late 1990s (15-year average) that is not seen in LUH1, RAMA and LUH2 reconstructions (between 1.2 and 1.6 Pg C a⁻¹ in this period).

Global average NPP was simulated to increase strongly over the last century due to the effect of higher vegetation productivity under an increased atmospheric CO₂ mixing ratio and (in cool areas) climate warming. Compared to reference simulations with LUC fixed in 1700 (red lines in Fig. 2d), all four LUC representations showed a reduced increase in NPP over the duration of the simulations (Table 3; Fig. S3), with the reduction due to LUC at the end of the historical period (averages 1998–2007) being 1.9 Pg C for LUH1, 2.4 Pg C for HYDE, 3.3 Pg C for RAMA and 2.0 Pg C for LUH2 LUC reconstructions (Table 3). The global total and the time series of NPP simulated with the four LUC reconstructions were very similar for RAMA and HYDE and were about 2 Pg C lower for LUH1 and LUH2 simulations for the entire simulation period (Fig. 2d), possibly as a result of a higher pasture area instead of natural (woody) vegetation in the LUH1 and LUH2 data in the extratropical regions of Africa and areas in eastern Europe.

For global C stocks, both in vegetation and soils, the positive trend induced by CO₂ fertilised vegetation growth (red lines in Fig. 2e and f) was counteracted by the effects of LUC change. A minimum of vegetation C stocks during the simulation period was simulated for all LUC reconstructions in the 1960s when LUC reduced vegetation C stocks on average by 108 Pg C (Table 3) compared to the reference simulations. Following the decline in conversion rates of natural into managed land thereafter, vegetation C stocks increased in response to vegetation productivity responding to the fertilising effect of increasing atmospheric CO₂ concentration. Vegetation C stocks at the beginning of the simulation period were again similar for RAMA and HYDE on the one hand and for LUH1 and LUH2 reconstructions on the other with about 482 Pg C but on average about 31 Pg C lower for LUH1 and LUH2 simulations because of the lower natural area in these datasets around 1700 (Table 3).

Time trends in soil C stocks over the simulation period followed similar trends as vegetation C stocks, albeit with a 5- to 10-year time lag (Fig. 2e, f). Loss in soil C as a result of accounting for LUC was a direct effect of the removal of biomass upon harvest, reducing the litter input in the following years, and the increase in soil C decomposition rates associated with tillage. On average, 73 Pg soil C was lost due to changes in LUC in the 2000s, with only a small variation of ± 1 Pg C between LUH1, RAMA and HYDE reconstructions and about 7 Pg C less for LUH2 (Table 3; Table S3 in the Supplement). Overall soil C stocks were again more sim-

Table 4. Changes in C stocks and fluxes in reconstructions of net and gross LUC changes for Europe (EU27 + CH). Values are always given as 10-year averages (except LUC areas and cumulative flux).

	Averaging period	Calculation	Unit	HILDA (net)	LUH1 (net)	Average and uncertainty* for the LUC models	HILDA (gross)	difference HILDA (gross–net)
Natural area	1900	A_{nat}	10^6 km^2	1.49	2.16	1.83 ± 0.47	1.49	0
Natural area	2010	A_{nat}	10^6 km^2	2.06	2.79	2.43 ± 0.52	2.06	0
Pasture area	1900	A_{pas}	10^6 km^2	1.62	0.97	1.30 ± 0.46	1.62	0
Pasture area	2010	A_{pas}	10^6 km^2	1.35	0.68	1.02 ± 0.47	1.35	0
Cropland area	1900	A_{crop}	10^6 km^2	1.59	1.57	1.58 ± 0.01	1.59	0
Cropland area	2010	A_{crop}	10^6 km^2	1.29	1.23	1.26 ± 0.04	1.29	0
Total change in natural area	1900–2010	A_{nat}	10^6 km^2	+0.57	+0.63	$+0.60 \pm 0.04$	+0.57	0
Total change in pasture area	1900–2010	A_{pas}	10^6 km^2	–0.29	–0.34	$–0.28 \pm 0.01$	–0.29	0
Total change in cropland area	1900–2010	A_{crop}	10^6 km^2	–0.28	–0.29	$–0.32 \pm 0.03$	–0.28	0
Total area under transition	1900–2010	ΣA_{trans}	10^6 km^2	1.47	1.87	1.67 ± 0.29	2.64	+1.17
E_{LUC}	1900–1909	$E_{LUCNet/Gross}$	Tg C a^{-1}	19	38	29 ± 13	29	+9
	1980–1989	$E_{LUCNet/Gross}$	Tg C a^{-1}	–25	–78	$–51 \pm 37$	–26	–1
	1990–1999	$E_{LUCNet/Gross}$	Tg C a^{-1}	–38	–84	$–61 \pm 33$	–38	0
	2001–2010	$E_{LUCNet/Gross}$	Tg C a^{-1}	–52	–80	$–66 \pm 20$	–51	+1
Cumulative E_{LUC} from 1900	1951–1960	$\Sigma E_{LUCNet/Gross}$	Tg C	586	1338	962 ± 532	915	+329
	2010	$\Sigma E_{LUCNet/Gross}$	Tg C	–936	–1890	674 ± 48	–531	+406
NPP	1900–1909	$NPP_{Net/Gross}$	Tg C a^{-1}	1464	1517	1490 ± 38	1462	–2
	2001–2010	$NPP_{Net/Gross}$	Tg C a^{-1}	2261	2361	2311 ± 71	2243	–18
Change in NPP due to LUC	1900–2010	$NPP_{Net/Gross} - NPP_{Ref}$	Tg C a^{-1}	–30	–10	$–20 \pm 14$	–44	–14
	2001–2010	$NPP_{Net/Gross} - NPP_{Ref}$	Tg C a^{-1}	–57	+10	$–23 \pm 47$	–74	–18
VegC	1900–1909	$VegC_{Net/Gross}$	Tg C	7755	9634	8694 ± 1328	7715	–40
	2001–2010	$VegC_{Net/Gross}$	Tg C	10518	13484	12001 ± 2097	10360	–159
Change in vegetation C due to LUC	1900–2010	$VegC_{Net/Gross} - VegC_{Ref}$	Tg C	–58	–234	$–146 \pm 125$	–199	–141
	2001–2010	$VegC_{Net/Gross} - VegC_{Ref}$	Tg C	+709	+1217	$+963 \pm 359$	+551	–159
Soil C	1900–1909	$SoilC_{Net/Gross}$	Tg C	58786	60672	59729 ± 1334	58775	–11
	2001–2010	$SoilC_{Net/Gross}$	Tg C	60016	62188	59761 ± 1536	59761	–254
Change in soil C due to LUC	1900–2010	$SoilC_{Net/Gross} - SoilC_{Ref}$	Tg C	–199	–314	$–256 \pm 81$	–368	–169
	2001–2010	$SoilC_{Net/Gross} - SoilC_{Ref}$	Tg C	–29	+291	$+131 \pm 226$	–283	–254

* Note that the uncertainty given here is calculated between two values.

ilar for RAMA and HYDE (average 1514 Pg C) and about 64 Pg C lower for LUH1 and LUH2 at the beginning of the simulation period (1700–1709; Fig. 2e; Table 3).

In Europe, LUC caused the emission of C until the 1960s but turned into a sink thereafter (Fig. 3a, negative E_{LUC}) as a result of the reduction in pastures and also croplands in the second half of the last century and the regrowth of natural (woody) vegetation (Fig. 1b). This development is shown in simulations with both HILDA and LUH1; however, the magnitude of the effect of LUC on C stocks and fluxes was somewhat stronger in simulations applying the LUH1 dataset, due to a higher deforestation rate in LUH1 until the 1950s (Fig. 3c; absolute land transitions were similar for both LUC datasets, Fig. 1d). In this sense, E_{LUC} decreases from about 19 Tg C a^{-1} (HILDA) and 38 Tg C a^{-1} (LUH1) in the 1900s to about $–52 \text{ Tg C a}^{-1}$ for HILDA and $–80 \text{ Tg C a}^{-1}$ for LUH1 in the 2000s, and cumulative E_{LUC} from 1900 to 2010 was $–936 \text{ Tg C}$ for HILDA and $–1890 \text{ Tg C}$ for LUH1 (Table 4).

As in the global simulations, a positive trend in NPP from 1900 was also simulated for Europe, which is linked to increasing CO_2 fertilisation (Fig. 3d). Accounting for changes in LUC reduced NPP in simulations applying HILDA by 57 Tg C a^{-1} but only slightly changed NPP in the 2000s

when applying LUH1 (increase of 10 Tg C a^{-1} , Table 4) because of highly productive croplands in central Europe (Fig. S4a). NPP simulated with HILDA was between 50 and 100 Tg C a^{-1} lower over the simulation period than simulated with LUH1 (Table 4). This also resulted from the lower share of natural areas in HILDA, as opposed to pastures, and thus lower productivity.

Similar trends in vegetation and soil C were simulated with both datasets, with changes dominated by deforestation over the first decades and reforestation thereafter (see Fig. 1b). In the 2000s vegetation C stocks were even higher under net LUC changes compared to the respective reference simulations (Table 4). C stocks in the vegetation of simulations using LUH1 were on average about 2000 Tg C and 25 % higher than simulations applying HILDA (Fig. 3e, Table 4). Differences in soil C stocks between the two LUC representations were small, with soil C being about 1900 Tg C higher in LUH1 simulations (3 % of soil C stocks projected with HILDA) at the beginning of the simulation period (Table 4). In comparison to the trend in C stored in vegetation, stocks in soils only increased from the 1950s on. Effects of LUC on C stocks in vegetation and soils were stronger for simulations applying LUH1 (Fig. 3e), showing increases in C stocks

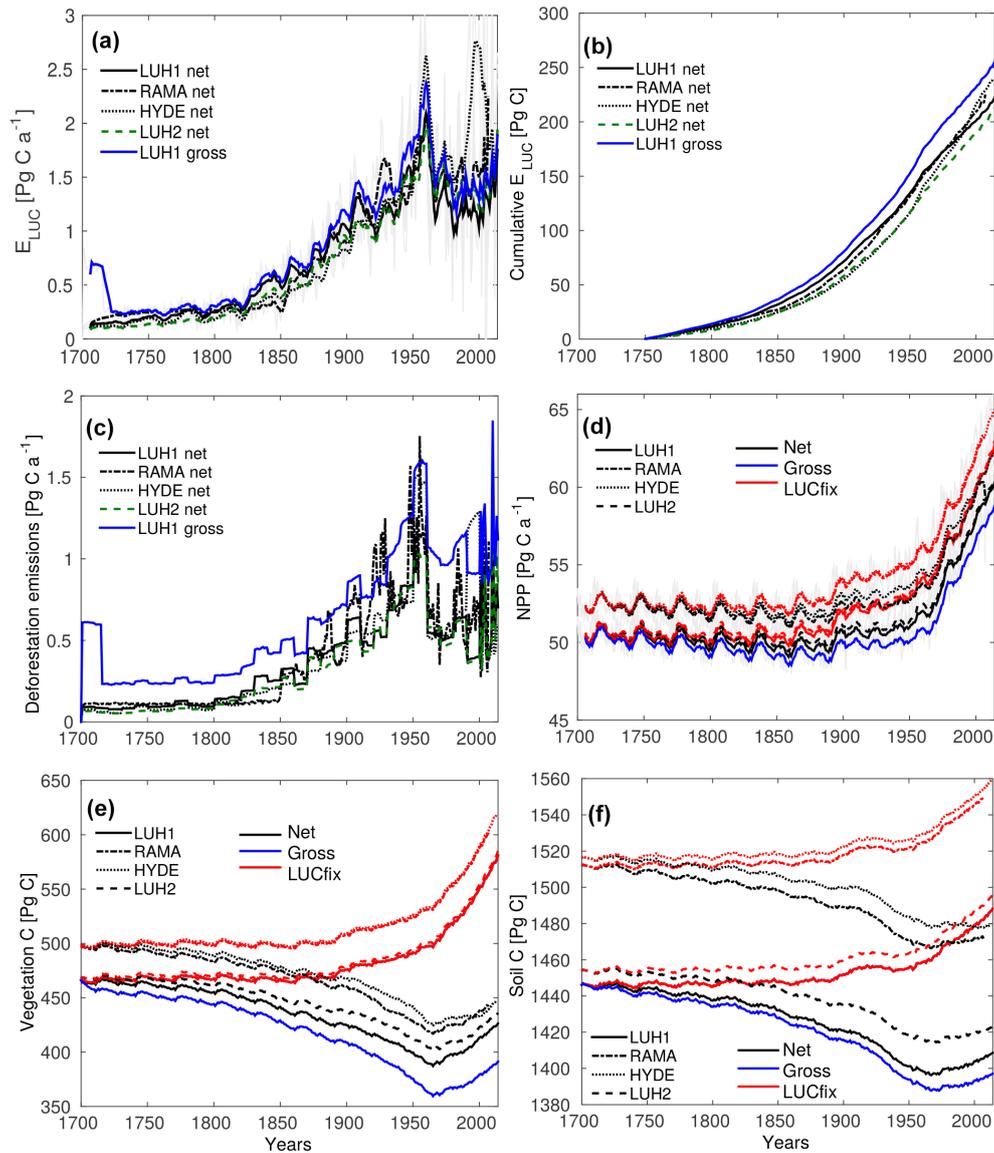


Figure 2. Effects of different land-use representations on global ecosystem C stocks and fluxes: land-use flux (a), cumulative land-use flux (b), deforestation emissions (c), net primary productivity (NPP) (d), vegetation (e) and soil carbon stocks (f). Flux values in (a) and (d) are given as 15-year averages with original values in the background. NPP, vegetation and soil C is shown for simulations experiencing net (or gross) LUC and in addition for simulations where LUC was kept at 1700 levels (see the “Methods” section; red lines in d, e, f).

in both central and eastern Europe but decreases in southern countries (Fig. S4b, c).

3.3 Global and European effects of accounting for gross land transitions

The global land area undergoing LUC when considering gross land transitions based on the LUH1 dataset (Fig. 1c) was 4.7 times the net area converted (total transitions 1700–2014, Table 3), with all additional land transitions in this dataset being generated by shifting cultivation in certain tropical regions (Fig. S1a). This increased the global land-use

flux E_{LUC} by about 0.14 to 1.64 Pg C a^{-1} at the end of the historical period (2005–2014 average flux) and resulted in cumulative E_{LUC} being 33 Pg C higher in 2014 for gross compared to net LUC simulations (Fig. 2a, b; Table 3). Global NPP was 1.5 Pg C a^{-1} (i.e. 3%) lower in simulations of gross land changes compared to the net simulations (Fig. 2d, Table 3), which was an effect of lower mean levels of forest canopy closure in the tropical areas subject to shifting cultivation. For the same reason, the reduction in vegetation C stocks as an effect of accounting for gross effects was high, with 35 Pg C , i.e. –8%, at the end of the simulation period. For soils, however, the effect was relatively

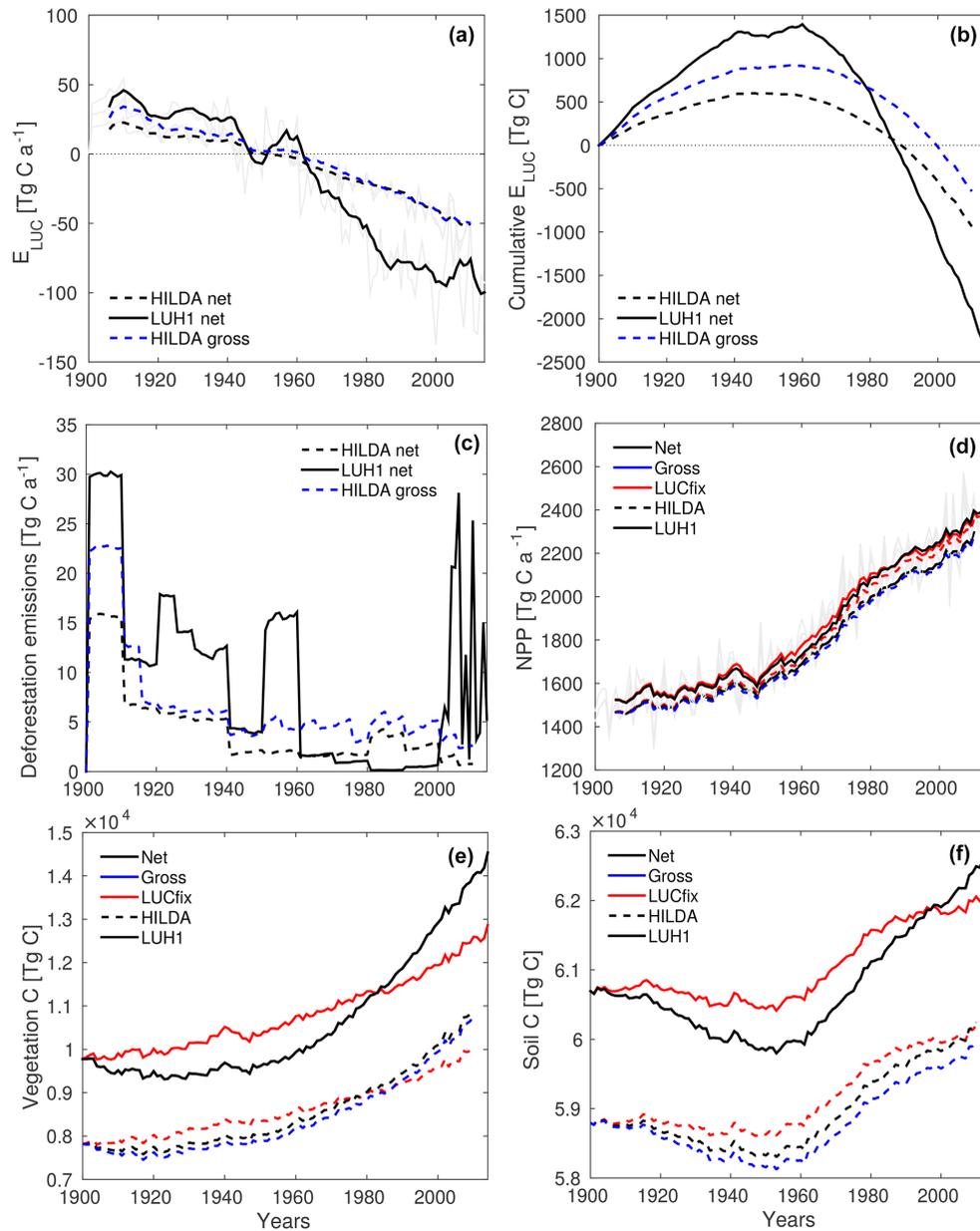


Figure 3. Effects of different land-use representations on ecosystem C stocks and fluxes for Europe (EU27 + CH): land-use flux (a), cumulative land-use flux (b), deforestation emissions (c), net primary productivity (NPP) (d), vegetation (e) and soil C (f). Flux values in (a) and (d) are given as 15-year averages with original values in the background. NPP, vegetation and soil C is shown for simulations experiencing net (or gross) LUC and in addition for simulations where LUC was kept at 1700 levels (see the “Methods” section; red lines in d, e, f).

low compared to the absolute size of soil C stocks, with an 11 Pg C reduction (-0.8% , 2005–2014; Fig. 2e and f; Table 3). The reduction in vegetation C stocks by the effects of LUC changes further increased by 24 % when accounting for gross LUC and for soil C stocks by 14 % (2005–2014, Table 3).

For Europe, the HILDA gross dataset predicted land transitions (Fig. 1d) that were about 1.4 times higher when accounting for gross transitions relative to net LUC changes

(total transitions 1900–2010, Table 4) (see also Fuchs et al., 2015b) with significant gross land changes occurring all over Europe (Fig. S1b). As a result, gross E_{LUC} was enhanced by about 11 Tg C a^{-1} in the beginning of the simulation period (1901–1910) compared to net E_{LUC} . Cumulative gross E_{LUC} was -531 Tg C in 2010, or 406 Tg C higher than E_{LUC} under net LUC transitions (-936 Tg C), representing a reduced cumulative sink as the result of higher previous emissions from LUC when considering gross transitions (Fig. 3b; Table 4).

NPP was also lower in gross simulations; however, differences were small (-18 Tg C a^{-1} ; see Table 4) and the gross simulation followed the same trend as the net LUC simulation. For C stocks on the European level, the difference between net and gross simulations was -158 Tg C for vegetation carbon and -254 Tg C for soil C stocks at the end of the simulation period (2001–2010; Fig. 3e and f; Table 4).

4 Discussion

4.1 Uncertainties in carbon stocks and fluxes due to the choice of historical LUC reconstruction

Resulting from the fact that historical reconstructions of land use and its changes are inherently uncertain because of the limited existing data base that needs complementary assumptions (e.g. on land rotation times), it is widely acknowledged that a key uncertainty in estimating changes in C stocks and fluxes as a response to LUC change stems from the choice of the LUC dataset (e.g. Houghton et al., 2012; Jain et al., 2013). With a detailed representation of succession when natural vegetation recolonises abandoned agricultural lands, the representation of croplands by a number of crop functional types and the consideration of C–N interaction in natural vegetation and crops, the LPJ-GUESS model considers key processes and interactions that are crucial when accurate estimates of C stocks and fluxes are derived based on detailed dynamics of LUC (see, e.g., Hickler et al., 2004; Lindeskog et al., 2013; Olin et al., 2015; Pugh et al., 2015; Zaehle, 2013).

Uncertainties in E_{LUC} resulting from the choice of LUC reconstruction (expressed as 1 standard deviation around decadal means), as quantified with the LPJ-GUESS model and the four global net LUC datasets, were ± 0.19 , ± 0.66 and $\pm 0.47 \text{ Pg C a}^{-1}$ (14, 39 and 29 % of E_{LUC}) in the 1980s, 1990s and 2000s respectively (see Table 3 for exact periods). Among the four datasets, before the 1960s, E_{LUC} using LUH1 and RAMA was similar, with the HYDE and LUH2 datasets differing somewhat from this as a result of regional differences in croplands and pastures, although global totals remained similar (e.g. HYDE showed less croplands in the northeastern US around 1900 than LUH1 or RAMA; LUH2 showed a similar cropland distribution to LUH1 but big differences in pastures in the southern parts of Russia; and LUH2 showed less pastures in southern Argentina and Kazakhstan than RAMA). These uncertainty calculations are consistent with estimates from previous studies in which a subset of the three LUC hindcasts (LUH1, RAMA and HYDE, partially as earlier versions) were applied, sometimes in combination with a bookkeeping method (Table S1). Uncertainties were also estimated for the 1980s as $\pm 0.30 \text{ Pg C a}^{-1}$ from a synthesis experiment of Houghton et al. (2012) and as $\pm 0.20 \text{ Pg C a}^{-1}$ for the 1990s as determined by Shevliakova et al. (2009) when using HYDE and RAMA cropland data (in earlier versions including wood harvest). For the 2000s, Jain

et al. (2013) found an uncertainty of $\pm 0.21 \text{ Pg C a}^{-1}$ when quantifying E_{LUC} with HYDE and RAMA datasets with both a coupled C–N DGVM and the bookkeeping approach from Houghton (2008). Our uncertainty of E_{LUC} due to the choice of LUC reconstruction was higher in the 1990s, where E_{LUC} derived using HYDE data was significantly higher due to a strong increase in pastures (Fig. 2a; Table S1). The uncertainty in cumulative E_{LUC} accumulated to $\pm 14 \text{ Pg C}$ for 1750–2007. For vegetation and soil carbon, the uncertainty introduced due to the choice of LUC reconstruction was ± 11 and $\pm 37 \text{ Pg C}$, respectively, in 1998–2007, translating into a change of 3 % of the average size of both vegetation and soil C stocks (Table 3; see Fig. S3 for regional differences for the entire simulation period). These uncertainties for vegetation C stocks are higher compared to the ones found by Arora and Boer (2010), who used only two of the LUC models applied here, and about the same size for C stocks in soils. This implies non-linear interactions between the DGVM structure and LUC dataset. We would expect uncertainty in C stocks and fluxes to increase, at least during the pre-1900 period, if LUC reconstructions applying a non-linear development of per capita land use were also considered, such as the KK10 dataset does for the period 8000 BC to AD 1850 (Kaplan et al., 2010).

For Europe, the uncertainty in E_{LUC} is also large with ± 37 , ± 33 and $\pm 20 \text{ Tg C a}^{-1}$ for the 1980s, 1990s and 2000s, values which are between 30 and 72 % of the average flux (Table 4). Differences result mainly from a disagreement in the amount of pastures between the LUC reconstructions, where the comparison is impaired by different definitions of the pasture class (see the “Methods” section). This highlights the problem of fundamentally different structures and assumptions between LUC models and DGVMs, recently identified as a major uncertainty in model estimates of E_{LUC} (Pongratz et al., 2014). Although forests, natural grasslands and pastures can show similar gross primary productivity (GPP), they significantly vary in their C sequestration potential in vegetation and soils also depending on their location within Europe (Ciais et al., 2008; Schulze et al., 2010). For instance, larger pasture areas in HILDA in southern Europe compared to LUH1 lead to an increase in vegetation C under LUC, whereas a decrease was simulated with LUH1 (Fig. S4). The productivity and carbon dynamics of croplands in LPJ-GUESS is mainly governed by the crop selection, the bioclimatic conditions of the land where the crop is planted and the degree of fertilisation and irrigation (for productivity of croplands under different degrees of fertilisation see also Ciais et al., 2010; Schulze et al., 2010). For instance, in Poland, high cropland fractions in LUH1 that were only marginally fertilised compared to croplands in western Europe decreased NPP under LUC, but increased vegetation and soil C (Fig. S4). Differences between the LUC reconstructions and therefore uncertainties in C stocks and fluxes converge during the first half of the 19th century (maps not shown). Vegetation C stocks derived with HILDA are lower

than estimates of Fuchs et al. (2015a; R. Fuchs, personal communication, 2016), using the same datasets (minor difference in net data; see the “Methods” section) and a bookkeeping method, but are within the uncertainty spanned by using HILDA and LUH1 net LUC datasets (Table S2).

It is important to consider that the relative uncertainties in LUC transitions between datasets are not constant through time, although the absolute uncertainties remain remarkably constant over the simulation period. The relative deviation of pasture area for the three global datasets was about $\pm 29\%$ before the 1850s, decreasing to about $\pm 9\%$ in the 2000s (Fig. S5c). Global cropland areas had a deviation of about $\pm 11\%$ in the 1700s, decreasing to below $\pm 2\%$ in the 2000s (Fig. S5c). For Europe the agreement of the two LUC reconstructions is high for croplands (average deviation $\pm 2\%$ for 1900–2010) but lower for pastures and natural areas (1900–2010 on average $\pm 36\%$ for pastures and $\pm 21\%$ for natural areas), with the deviation increasing until 2010 for pastures (Fig. S5d). The general agreement on the fractional coverage of natural land, pastures and croplands is higher for the periods after 1960, when FAO statistical data and later improved data from satellites became available (see also Houghton, 2010; Verburg et al., 2011). Before this period, the extrapolation of historical LUC information was very much dependent on the applied model algorithms in combination with census data. LUC reconstructions also differ in the resolution of past LUC changes, providing annual time steps (RAMA for the entire historical period; LUH1, HYDE, LUH2 after 2000) or originating from decadal aggregations (HILDA for the entire historical period; LUH1, HYDE, LUH2 until 2000). Such methodological discrepancies and artefacts from the LUC modelling significantly overlay the observed trends in the LUC reconstructions (compare Fig. 1c, d) and are included in simulated C stocks and fluxes (Figs. 2, 3). It is noted that LUH1, HYDE and LUH2 net LUC data have LUC features in common and cannot be regarded as completely independent datasets, as LUH data are generally built based on HYDE input data. Here, two subsequent (HYDE as HYDE3.1.1 and LUH2 as based on HYDE3.2) and one intermediate (LUH1 as based on an early version of HYDE3.2) developments of LUH and HYDE were used (see the “Methods” section); however, no versions directly building up on each other were considered (e.g. dataset 1 is a version of HYDE that was used as direct basis for a version of LUH that was considered as dataset 2).

4.2 Uncertainties in carbon stocks and fluxes due to accounting for gross land transitions

The consideration of detailed gross land conversions in our simulations increased the effects of LUC on carbon storage and fluxes due to larger areas transitioning between land-use types (Fig. 1c, d). The increase of about 16% in average net annual E_{LUC} and 15% in cumulative E_{LUC} (Table 3; change due to gross relative to net, 1750–2014) compared well with

previous estimates (Table S3). The effect of shifting cultivation on cumulative E_{LUC} was quantified by Olofsson and Hickler (2008) by using the LPJ model (Sitch et al., 2003), which has similarities in the way plant and soil physiological processes are calculated to the model used here but has a simpler representation of vegetation dynamics and croplands and no C–N dynamics. Olofsson and Hickler (2008) found an increase in cumulative E_{LUC} by 28 and 29% for 1700–1990 and 1850–1990, whilst Stocker et al. (2014), using a model with coupled C–N dynamics at a $1^\circ \times 1^\circ$ spatial resolution, reported an increase by 15% (despite the coarser spatial resolution acting to accentuate the gross–net differences). Using a bookkeeping model, Hansis et al. (2015) found an enhancement of E_{LUC} by 22–24% over the period 1500–2012 at a $0.5^\circ \times 0.5^\circ$ spatial resolution. The combined effect of shifting cultivation and wood harvest on cumulative E_{LUC} was summarised by Houghton et al. (2012) as an increase by 25–35%, and Wilkenskjeld et al. (2014) found an increase in cumulative E_{LUC} of 61% (51% without the effect of wood harvest). Shevliakova et al. (2013) provide an estimate of E_{LUC} under gross transitions including wood harvest for the period 1860–2005 using a combination of modelled C fluxes and a bookkeeping method to derive E_{LUC} that is fairly close to the value calculated in this study. For total land C stocks, Shevliakova et al. (2009) reported an increase in E_{LUC} of 49% due to shifting cultivation and wood harvest and concluded from this that the effect on land carbon losses was comparable in magnitude to the effect of cropland and pasture expansion. In our study we quantified an increase in E_{LUC} of 39% total C globally due to shifting cultivation alone (Table S3). Of these studies, only the model of Stocker et al. (2014) (Table S3), accounts for C–N interactions. C–N interactions have previously been found to enhance LUC emissions (Jain et al., 2013). In addition, all these studies differ from our study in the DGVM used, the LUC datasets and climate model data applied, and the process representations in the models. All studies except Olofsson and Hickler (2008) applied spatial resolutions coarser than the 0.5° applied here (1° or $\sim 2^\circ$; see Table S3).

In our global experiment, the only contribution of gross land changes came from shifting cultivation in certain tropical areas. These gross changes were implemented in the LUH1 dataset based on assumed spatial extension and transition rates and were reflected in significantly increased rates of deforestation and reforestation in gross simulations (Fig. 4a). As would be expected, removing forest material from the system through harvest or the burning of cleared vegetation, instead of it entering the soil pool through litter decomposition, reduces soil C content. However, the soil C losses are much less marked than those in vegetation, perhaps reflecting a dominance of vegetation carbon turnover by leaves and fine roots in LPJ-GUESS, which are inputs of C to the litter pool which are less affected by the harvesting of vegetation on multi-annual rotation periods than woody inputs.

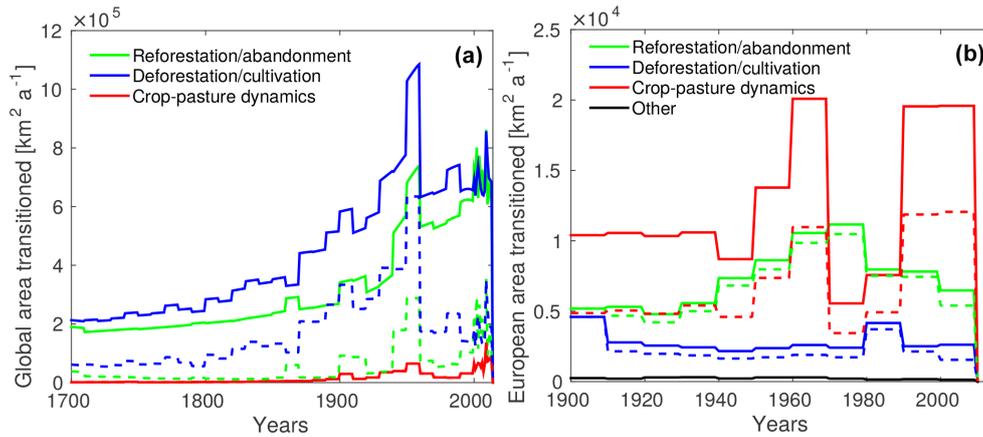


Figure 4. Annual area transitions for major land change processes for global (a, data from LUH1) and European (b, EU27 + CH, data from HILDA) gross (solid lines) and net (dashed lines) datasets. The class “other” in (b) includes transitions between natural vegetation and barren land represented in the HILDA dataset.

In comparison to the global gross land transitions, which accounted only for shifting cultivation with uniform assumptions regarding cultivation cycles, the European gross dataset accounted for irregular land-use and land-cover changes based on regionally available empirical evidence (national statistics and maps). E_{LUC} when accounting for gross land changes was about 53 % higher in the first simulation decade and converged to minor differences from about 1980 onwards (Fig. 3a). Since LUC in Europe developed from being a source of C in the beginning of the 1900s to being a sink after about 1960, the small difference in C stocks and fluxes as a result of accounting for gross transitions, in addition to net land changes, delayed the year in which cumulated E_{LUC} switched from a source to a sink. Thus, the sink capacity of cumulative E_{LUC} in the 2000s was reduced, as were increases in vegetation and soil C stocks (Fig. 3b, e, f). To our knowledge, no previous studies are available in which E_{LUC} for Europe was derived under net and gross LUC. The effect of accounting for gross land transitions on vegetation C was negative in our simulations, with vegetation C under gross LUC about 2 % lower than under net LUC, because the increased number of regrowing stands under higher land transitions lowered mean forest canopy closure, which also lowered NPP and, ultimately, soil C (see Table 4). In contrast Fuchs et al. (2015a and unpublished results), using a bookkeeping method, derived about 1 % higher vegetation C stocks under gross LUC for the same area (Table S2). Discrepancies result from major methodological differences between the bookkeeping and process-based approach and also from Fuchs et al. (2015a) not accounting for C stocks in croplands and pastures.

Gross LUC transitions in Europe over the entire simulation period were dominated by conversions between pastures and croplands, i.e. cropland expansion into pasture areas and abandonment of croplands and their conversion into

pastures (Fig. 4b), that were direct adjustments to market demands and changes in land-use-related policies. Apart from this, LUC in Europe was characterised by the abandonment of agricultural land and reforestation peaking in the 1970s. Reforestation of European grasslands was reported to entail a reduction in soil C stocks and an increase in vegetation C (Schulze et al., 2010; Fig. 3e, f) and therefore a positive E_{LUC} (Fig. 3a). After a first period of regrowth, the additional tree biomass and increased litter inputs to soils balanced soil C losses, so that vegetation and soil stocks increased in the second half of the 20th century because wood harvest was lower than growth (e.g. Ciaia et al., 2008), thereby contributing to the LUC sink capacity. Because land abandonment and reforestation are one-directional LUC changes which are represented in the same way in net and gross HILDA data (see Fuchs et al., 2015b), this did not lead to major differences between net and gross simulations. It should be noted that with its current four LUC classes, the LPJ-GUESS model was not able to make full use of the HILDA LUC dataset, as not all HILDA land-cover classes were represented, e.g. urbanisation (urban areas were assigned to the barren LUC class; see the “Methods” section), causing $\sim 18\%$ of gross land-use changes between 1900 and 2010 (Fuchs et al., 2015b).

4.3 Uncertainties in the modelling approach

The effects and uncertainties discussed above must be considered in relation to other uncertainties arising in the modelling process (e.g. different model implementations in respect of the representation of LUC and changes therein, treatment of environmental change). A meta-analysis by Houghton et al. (2012) estimated the uncertainty in E_{LUC} arising from the applied modelling approach and method to be in the range of $\pm 0.2 \text{ Pg C a}^{-1}$ and that due to data-related uncertainty and incomplete process understanding to be in the range of $\pm 0.5 \text{ Pg C a}^{-1}$. The complex linkages between

the contributing factors, however, made it difficult to attribute uncertainties to their sources (see also Jain et al., 2013). A consideration of C–N interaction in vegetation and soils, as was done in this study, is important when studying the effects of environmental drivers such as LUC on carbon emissions; however, with the exception of a very few models (e.g. Jain et al., 2013; Smith et al., 2014; Xu-Ri and Prentice, 2008; Zaehle and Friend, 2010), most models do not represent N cycling. We did not quantify the effect of C–N interactions in this study, but we note that our estimates of cumulative E_{LUC} from 1850 to 2005 with net HYDE (Table S3) are about 2 % lower than those of Pugh et al. (2015) using a version of LPJ-GUESS without C–N interactions and the same LUC data. This is in opposition to the findings of Jain et al. (2013), who found about 40 % higher E_{LUC} globally when accounting for N dynamics and N limitation.

The implementation of gross transitions in a DGVM framework is subject to considerable uncertainty. By varying the minimum age upon which a regrowing natural stand becomes eligible for clearance again between 5 and 30 years (recalling that natural stands are removed in order of age when natural vegetation is reduced; see the “Methods” section), cumulative E_{LUC} was found to differ by ± 20 Pg C (10 %) (1900–2014, LUH1 only; results not shown). Hence, choosing too young a clearance age would lead to considerable underestimates of E_{LUC} which highlights an important, and hitherto unremarked upon, implementation uncertainty for including gross transitions in DGVM simulations. In the same way, other land-use and land-cover-change-related processes (fate of harvested wood, residue management, occurrence of disturbances, including fire, etc.) might differ under repeated land transitions such as shifting cultivation, and the realistic representation of these interactions in process-based models remains a subject for further research. Wood harvest mainly in extratropical regions was assessed to account for an increase in E_{LUC} of 0.2–0.3 Pg C a⁻¹ (Houghton et al., 2012). We did not account for wood harvest in this study as the uncertainty in the actual spatial pattern of wood harvest in combination with the ways wood harvest is done in practice over the globe (clear cut, selective harvesting of specific age classes or a mixture of both) introduces many possibilities as to how this process can be implemented in DGVMs. In a model such as LPJ-GUESS, where forest ecosystem and wood parameters vary significantly over tree age classes, this would result in a wide span of possible solutions depending on the parameters used for the implementation of wood harvest that would be better addressed in a dedicated sensitivity study investigating a variety of possible implementations rather than with a single representation.

Simulation results of biogeochemical cycles with a DGVM such as LPJ-GUESS depend critically on the year when simulations are started. In this study we tested the effect of starting LUH1 simulations in either 1700 or 1900 (Fig. S6). E_{LUC} cumulated over 1950–2014 was 17 % lower and soil C was about 33 Pg C (2 %) lower when simulations

were started in 1900, while vegetation C stocks were similar (net LUC, Table S4). This emphasises the impact on simulation results of soil legacy emissions resulting from previous LUC changes (see, e.g., Gasser and Ciais, 2013, and Sentman et al., 2011, for legacy fluxes, and Pugh et al., 2015, for breakdown of LUC emissions). Hansis et al. (2015) in contrast found a 28 % higher E_{LUC} over the period 2008–2012 when legacy emissions predating 2008 were excluded. This resulted from previous LUC transitions generating more uptake from regrowth than loss from soil decomposition during the 2008–2012 period, and it further exemplifies the considerable importance of legacy effects for the calculation of instantaneous emissions. They also emphasise that the change in soil and vegetation stocks induced by previous LUC can substantially modify E_{LUC} calculations compared to an assumption of equilibrium in the initial stocks. Often such changes in initial conditions tend to lower vegetation C stocks and thus subsequent deforestation emissions. Our 17 % lower E_{LUC} excluding legacy emissions accounts for both these effects. To exclude this effect from the analysis of uncertainties due to LUC dataset selection and the effect of accounting for gross LUC transitions, all global simulations were started in 1700. For Europe, where HILDA reconstructions were not available before 1900, simulations were started in 1900 for both HILDA and LUH1 to ensure comparability.

5 Conclusions

Global and European carbon stocks and fluxes and the effects of changes in LUC were shown to be subject to significant uncertainties resulting from the choice of historical LUC reconstruction. In our global simulations, data from HYDE and RAMA and from LUH1 and LUH2 often lead to similar results in ecosystem C stocks and fluxes. Therefore, LUH1 and LUH2 as the more recent developments of the four considered reconstructions (both based on HYDE version 3.2; however, LUH1 is based on an intermediate version and LUH2 on the final version) differ more from older developments than those do from each other (see the “Methods” section for model versions used in this study). For Europe, variables predicted based on the newly available HILDA dataset were similar to those resulting from using LUH1 for Europe; however, LUH1 predicted larger changes under LUC. Differences in the effects of both global and European LUC on C stocks and fluxes were found to be mainly based on the total area and spatial distribution of pastures in the datasets; however, note that the area of pastures is impaired by different classifications used by the LUC models. To account for the uncertainty arising from different reconstructions of historical LUC in the dynamic modelling of C stocks and fluxes and to provide realistic estimates of this uncertainty for the land-use C flux, the consideration of multiple LUC reconstructions exploring the full range of reasonable assumptions

is needed, as well as efforts to narrow the uncertainty in constructions of historical land use. Multiple LUC reconstructions were calculated by Hurtt et al. (2011), but the consequences of uncertainty in land-use transitions are not routinely explored by the carbon cycle community. This goes along with the reduction in uncertainties in the implementation of these datasets and different forms of LUC in DGVMs which has recently been the focus of discussion (Pongratz et al., 2014; Pugh et al., 2015; Stocker and Joos, 2015).

The results herein show that considering gross land conversions significantly increased the effect of LUC change on C stocks and fluxes. Most noticeably, the land-use C flux was enhanced by about 15 % of carbon released in addition to when only accounting for net land changes (cumulated 1750–2014), primarily resulting from a reduction in vegetation C storage. Note that for DGVMs operating at a lower spatial resolution than the $0.5^\circ \times 0.5^\circ$ used here, the underestimation of E_{LUC} would be even larger, as shown by Wilkenskjaeld et al. (2014). Given the large percentage enhancements in E_{LUC} found by considering gross transitions, this should be the preferred method whenever possible.

The implementation of gross land transitions, however, poses technical and parameterisation challenges to the process-based models. It also relies on extensive information on historical land-use transitions, which is largely lacking; at present, only a few LUC models are able to represent gross land changes on larger spatial scales, providing a limited basis to characterise the uncertainty. The LUH datasets are the only global-scale reconstructions representing gross land changes by explicitly implementing shifting cultivation in certain tropical areas with assuming a fixed period of 15 years for which land is cultivated before abandonment (for LUH1). For Europe, the HILDA dataset is the first reconstruction representing gross land transitions which are based on actual LUC inventory data and complementary model assumptions. The reconstruction of detailed regional sub-grid land transitions, and possibly more realistic patterns of shifting cultivation today, is restricted by the lack of reliable information on continental and global-scale historical land transitions. New datasets based on archived LUC data and remote-sensing sources are currently becoming available with high spatial resolution for the global to continental scale (Chen et al., 2015; European Environment Agency, 2014; Wang et al., 2015) and on a national to regional level (Homer et al., 2015; MOFOR, 2016; RCMRD, 2016; Roy et al., 2015; TerraClass, 2016). New promising efforts also provide LUC change data globally derived from remote sensing with a 250 m spatial resolution (Wang et al., 2015) and with a 30 m spatial resolution (Chen et al., 2015). In the coming years, new high-resolution LUC datasets can be expected from the Landsat archives (<http://earthexplorer.usgs.gov>) and the new Sentinel missions (Aschbacher and Milagro-Pérez, 2012). These will contribute to closing the information gap and in this way improve the calibration of LUC models to represent the underlying processes, reduce the uncertainty in

ecosystem functions such as the present-day land-use flux, and provide enhanced information for, e.g., the assessment of ecosystem services and biodiversity indicators in the future.

6 Data availability

Researchers interested in the LPJ-GUESS source code can contact the model developers (<http://iis4.nateko.lu.se/lpj-guess/contact.html>, Smith et al., 2014). The HILDA data set can be downloaded freely at www.wur.eu/hilda (Fuchs et al., 2014). LPJ-GUESS simulation results are stored at KIT-IMK-IFU computing facilities and can be obtained on request (anita.bayer@kit.edu).

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Author contributions. A. D. Bayer, T. A. M. Pugh, and A. Arneth conceived and designed the experiments. M. Lindeskog realised the technical implementation of gross land-use changes in LPJ-GUESS. P. M. Anthoni processed LUH2 LUC data. A. D. Bayer carried out the model simulations and led the data analysis, with contributions from all authors. R. Fuchs contributed the HILDA land-use data. A. D. Bayer led the writing of the manuscript with contributions from all authors.

Competing interests. The authors declare that they have no conflict of interest.

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