Regional variation in the effectiveness of methane-based and land-based climate mitigation options

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Abstract. Scenarios avoiding global warming greater than 1.5 or 2°C, as stipulated in the Paris Agreement, may require the combined mitigation of anthropogenic greenhouse gas emissions alongside enhancing negative emissions through approaches such as afforestation/reforestation (AR) and biomass energy with carbon capture and storage (BECCS). We use the JULES land-surface model coupled to an inverted form of the IMOGEN climate emulator to investigate mitigation scenarios that achieve the 1.5 or 2°C warming targets of the Paris Agreement. Specifically, within this IMOGEN-JULES framework, we focus on and characterise the global and regional effectiveness of land-based (BECCS and/or AR) and anthropogenic methane (CH_4) emission mitigation, separately and in combination, on the anthropogenic fossil fuel carbon dioxide (CO_2) emission budgets (AFFEBs) to 2100. We use consistent data and socio-economic assumptions from the IMAGE integrated assessment model for the second Shared Socioeconomic Pathway (SSP2). The analysis includes the effects of the methane and carbon-climate feedbacks from wetlands and permafrost thaw, which we have shown previously to be significant constraints on the AFFEBs.

Globally, mitigation of anthropogenic CH_4 emissions has large impacts on the anthropogenic fossil fuel emission budgets, potentially offsetting (i.e. allowing extra) carbon dioxide emissions of 188-212 GtC. This is because of (a) the reduction in the direct and indirect radiative forcing of methane in response to the lower emissions and hence atmospheric concentration of methane; and (b) carbon-cycle changes leading to increased uptake by the land and ocean by CO_2-based fertilisation. Methane mitigation is beneficial everywhere, particularly for the major CH_4-emitting regions of India, USA and China. Land-based mitigation has the potential to offset 51-100 GtC globally, the large range reflecting assumptions and uncertainties associated with BECCS. The ranges for CH_4 reduction and BECCs implementation are valid for both the 1.5° and 2°C warming targets.

That is the mitigation potential of the CH_4 and of the land-based scenarios is similar for whether society aims for one or other of the final stabilised warming levels. Further, both the effectiveness and the preferred land-management strategy (i.e., AR or BECCS) have strong regional dependencies. Additional analysis shows extensive BECCS could adversely affect water security for several regions. Although the primary requirement remains mitigation of fossil fuel emissions, our results highlight the potential for the mitigation of CH_4 emissions to make the Paris climate targets more achievable.

1 Introduction

The stated aims of the Paris Agreement of the United Nations Framework Convention on Climate Change (UNFCCC, 2015) are "to hold the increase in global average temperature to well below 2°C and to pursue efforts to limit the increase to 1.5°C". The global average surface temperature for the decade 2006-2015 was 0.87°C above pre-industrial levels and is likely to reach 1.5°C between the years 2030 and 2052, if global warming continues at current rates (IPCC, 2018). The IPCC Special Report on Global Warming of 1.5°C (IPCC, 2018) gives the median remaining carbon budgets between 2018 and 2100 as 770 $GtCO_2$ (210 GtC) and 1690 $GtCO_2$ (~461 GtC) to limit global warming to 1.5°C and 2°C, respectively. These budgets represent ~20 and ~41 years at present-day emission rates. The actual budgets could however be smaller, as they exclude Earth system feedbacks such as CO_2 released by permafrost thaw or CH_4 released by wetlands. Meeting the Paris Agreement goals will, therefore, require sustained reductions in sources of fossil carbon emissions, other long-lived anthropogenic greenhouse gases (GHGs) and some short-lived climate forcers (SLCFs) such as methane (CH_4), alongside increasingly extensive implementations of carbon dioxide removal (CDR) technologies (IPCC, 2018). Accurate information is needed on the range and efficacy of options available to achieve this.

Biomass energy with carbon capture and storage (BECCS) and afforestation/reforestation (AR) are among the most widely considered CDR technologies in the climate and energy literature (Minx et al., 2018) . For scenarios consistent with a 2°C warming target, the review by Smith et al. (2016) finds this may require (1) a median removal of 3.3 GtC yr^{-1} from the atmosphere through BECCS by 2100 and (2) a mean CDR through AR of 1.1 GtC yr^{-1} by 2100, giving a total CDR equivalent to 47% of present-day emissions from fossil fuel and other industrial sources (Le Quéré et al., 2018). Although there are fewer scenarios that look specifically at the 1.5°C pathway, BECCS is still the major CDR approach (Rogelj et al., 2018). For the default assumptions in Fuss et al. (2018), BECCS would remove a median of 4 GtC yr^{-1} by 2100 and a total of 41-327 GtC from the atmosphere during the twenty-first century, equivalent to about 4-30 years of current annual emissions. The land requirements for BECCS will be greater for the 1.5°C target within a given shared socio-economic pathway (e.g., SSP2), although published estimates are similar for the two warming targets, with between 380-700 Mha required for the 2°C target (Smith et al., 2016) and greater than 600 Mha for the 1.5°C target (van Vuuren et al., 2018). This is because the land requirements for bioenergy production differ strongly across the different SSPs, depending on assumptions about the contribution of residues, assumed yields and yield improvement, start dates of implementation and the rates of deployment.

While the CDR figures assume optimism about the mitigation potential of BECCS, concerns have been raised about the potentially detrimental impacts of BECCS on food production, water availability and biodiversity, e.g., (Krause et al., 2017;Heck et al., 2018). Others note the risks and query the feasibility of large-scale deployment of BECCS e.g. (Anderson and Peters, 2016;Vaughan and Gough, 2016;Vaughan et al., 2018).

Harper et al. (2018) find the overall effectiveness of BECCS to be strongly dependent on the assumptions concerning yields, the use of initial above-ground biomass that is replaced and the calculated fossil-fuel emissions that are offset in the energy system. Notably, if BECCS involves replacing ecosystems that have higher carbon contents than energy crops, then AR and avoided deforestation can be more efficient than BECCS for atmospheric CO_2 removal over this century (Harper et al., 2018).

Mitigation of the anthropogenic emissions of non-CO_2 GHGs such as CH_4 and of SLCFs such as black carbon have been shown to be attractive strategies with the potential to reduce projected global mean warming by 0.22-0.5°C by 2050 (Shindell et al., 2012;Stohl et al., 2015). It should be noted that these were based on scenarios with continued use of fossil fuels. Through the link to tropospheric ozone (O_3), there are additional co-benefits of CH_4 mitigation for air quality, plant productivity and food production (Shindell et al., 2012) and carbon sequestration (Oliver et al., 2018). Control of anthropogenic CH_4 emissions leads to rapid decreases in its atmospheric concentration, with an approximately 9-year removal lifetime (and as such is an SLCF). Furthermore, many CH_4 mitigation options are inexpensive or even cost negative through the co-benefits achieved (Stohl et al., 2015), although expenditure becomes substantial at high levels of mitigation (Gernaat et al., 2015). The extra "allowable" carbon emissions from CH_4 mitigation can make a substantial difference to the feasibility or otherwise of achieving the Paris climate targets (Collins et al., 2018).

Some increases in atmospheric CH_4 are not related to direct anthropogenic activity, but indirectly to climate change triggering natural carbon and methane-climate feedbacks. These effects could act as positive feedbacks, and thus in the opposite direction to the mitigation of anthropogenic CH_4 sources. Wetlands are the largest natural source of CH_4 to the atmosphere and these emissions respond strongly to climate change (Melton et al., 2013;Gedney et al., 2019). A second natural feedback is from permafrost thaw. In a warming climate, the resulting microbial decomposition of previously frozen organic carbon is potentially one of the largest feedbacks from terrestrial ecosystems (Schuur et al., 2015). As the carbon and CH_4 climate feedbacks from natural wetlands and permafrost thaw could be substantial, this causes a reduction in anthropogenic CO_2 emission budgets compatible with climate change targets (Comyn-Platt et al., 2018;Gasser et al., 2018).

This paper models the potential for mitigation of greenhouse gases to contribute to meeting the Paris targets of limiting global warming to 1.5°C and 2°C respectively. Specifically, we investigate the effectiveness of mitigation of anthropogenic methane emissions and land-based mitigation (e.g., implementation of BECCS and AR), combining results from three recent papers (Collins et al., 2018;Comyn-Platt et al., 2018;Harper et al., 2018). We determine the effectiveness of these approaches in terms of their impact on the anthropogenic fossil fuel CO_2 emissions budget consistent with stabilising temperature at 1.5°C and 2° C of warming. The more effective the mitigation option, the larger the fossil fuel CO_2 emissions budget consistent with stabilisation at a given level. We estimate the impact of these mitigation scenarios relative to an existing scenario of greenhouse

gas concentrations (based on the IMAGE SSP2 baseline), spanning uncertainties in both climate model projections (both global warming and regional climate change), process representation and the efficacy of BECCS. Sect. 2 provides a brief description of the models, the experimental set-up and the key datasets used in the model runs and subsequent analysis. Sect. 3 presents and discusses the results, starting with a global perspective before addressing the regional dimension. For BECCS, we additionally investigate the sensitivity to key assumptions and consider the implications for water security. Sect. 4 contains our conclusions.

2 Approach and Methodology

Our overall modelling strategy is as follows. The starting point is the prescription of global temperature profiles that match the historical record, followed by a transition to a future stabilisation at either 1.5 or 2.0°C above pre-industrial levels. For these profiles, we then determine the related pathways in atmospheric radiative forcing by inversion of the global energy balance component of the IMOGEN impacts model. IMOGEN "Integrated Model Of Global Effects of climatic aNomalies" (Sect. 2.2) (Huntingford et al., 2010;Comyn-Platt et al., 2018) is an intermediate complexity climate model, which emulates 34 models in the CMIP5 climate model ensemble. Hence our radiative forcing (RF) trajectories have uncertainty bounds, reflecting the different thermal sensitivities of existing climate models.

For each radiative forcing pathway, we subtract the individual RF components for non-CO_2 and non-CH_4 radiatively-active gases that are perturbed by human activity, and for scenarios taken from the IMAGE integrated assessment model. Then, for CH_4, we use an understanding of its atmospheric lifetime to translate methane emissions into their atmospheric concentrations. The related RF for CH_4 is also subtracted from the overall value. Hence the remaining RF is that available for changes to atmospheric CO_2 concentration. The IMOGEN model uses pattern-scaling, again fitted to the same 34 climate models, to estimate local changes in near-surface meteorology. Combined with our global temperature pathways, these pattern-based changes (as well as atmospheric CO_2 concentration) drive the Joint UK Land-Environment Simulator land surface model (JULES, Sect. 2.1) (Best et al., 2011;Clark et al., 2011). JULES estimates atmosphere-land CO_2 exchange, and similarly, IMOGEN contains a single global description of oceanic CO_2 draw-down. These two estimates of carbon exchanges with the land and ocean respectively, in conjunction with atmospheric storage being linear in the CO_2 pathway, finally determine by simple summation compatible CO_2 emissions from fossil fuel burning. We call this the anthropogenic (CO_2) fossil fuel emission budgets (AFFEB) compatible with the warming pathway, subject to the assumptions made for non-CO_2 forcings.

Our numerical simulation structure allows us to investigate the implications of three different key changes on AFFEB, for stabilisation at both 1.5 and 2.0°C, and in a structure that captures features of a full set of climate models. First and maybe most importantly, we work to understand how regional reductions in CH_4 emissions allow higher values of AFFEB. Second, we consider how alternative scenarios of BECCS implementation alter atmosphere-land CO_2 exchanges, and again presented as the resultant implications for AFFEB. Third, we determine how the newer understanding of warming impacts on wetland

methane emissions also affects AFFEB. Figure 1 captures the modelling framework, derivation of AFFEB, and our numerical experiments in a single overall schematic diagram.

Each of the scenarios investigated using the IMOGEN-JULES framework comprises 2 ensembles of 136 members, one ensemble for each of the warming targets. We make use of these ensembles to derive an "uncertainty" in the derived carbon budgets, specifically from climate change (as given by the 34 CMIP5 models) and from key land-surface processes (methane emissions from wetlands and the ozone vegetation damage). The climate change uncertainty comprises both the range of climate sensitivities of the CMIP5 models and the different regional patterns in the models. We use the median of the 136-member ensemble as the central value to derive the carbon budgets and the interquartile range (25-75%) for the uncertainty.

2.1 The JULES model

We use the JULES land surface model (Best et al., 2011;Clark et al., 2011), release version 4.8, but with a number of additions required specifically for our analysis:

1. Land use: We adopt the approach used by Harper et al. (2018) and prescribe *managed* land-use and land-use change (LULUC). On land used for agriculture, C3 and C4 grasses are allowed to grow to represent crops and pasture. The land-use mask consists of an annual fraction of agricultural land in each grid cell. Historical LULUC is based on the HYDE 3.1 dataset (Klein Goldewijk et al., 2011), and future LULUC is based on two scenarios (SSP2 RCP-1.9 and SSP2 baseline), which were developed for use in the IMAGE integrated assessment model (IAM) (van Vuuren et al., 2017;Doelman et al., 2018) (see also Sect. 2.3).

   Natural vegetation is represented by nine plant functional types (PFTs): broadleaf deciduous trees, tropical broadleaf evergreen trees, temperate broadleaf evergreen trees, needle-leaf deciduous trees, needle-leaf evergreen trees, C3 and C4 grasses, deciduous and evergreen shrubs (Harper et al., 2016). These PFTs are in competition for space in the non-agricultural fraction of grid cells, based on the TRIFFID (Top-down Representation of Interactive Foliage and Flora Including Dynamics) dynamic vegetation module within JULES (Clark et al., 2011). A further four PFTs are used to represent agriculture (C3 and C4 crops, and C3 and C4 pasture), and harvest is calculated separately for food and bioenergy crops (see Sect. 2.4.3, where we describe the modelling of carbon removed via bioenergy with CCS). When natural vegetation is converted to managed agricultural land, the vegetation carbon removed is placed into woody product pools that decay at various rates back into the atmosphere (Jones et al., 2011). Hence, the carbon flux from LULUC is not lost from the system. There are also four non-vegetated surface types: urban, water, bare soil and ice.

2. Soil carbon: Following Comyn-Platt et al. (2018), we also use a 14 layered soil column for both hydro-thermal (Chadburn et al., 2015) and carbon dynamics (Burke et al., 2017a). Burke et al. (2017b) demonstrated that modelling the soil carbon fluxes as a multi-layered scheme improves estimates of soil carbon stocks and net ecosystem exchange. In addition to the vertically discretised respiration and litter input terms, the soil-carbon balance calculation also includes a diffusivity term to represent cryoturbation/bioturbation processes. The freeze-thaw process of

cryoturbation is particularly important in cold permafrost-type soils (Burke et al., 2017b). Following Burke et al. (2017a), we diagnose permafrost wherever the deepest soil layer is below 0°C (assuming that this layer is below the depth of zero annual amplitude, i.e., where seasonal changes in ground temperature are negligible (≤0.1 °C)). Further, for permafrost regions, there is an additional variable to trace or diagnose "old" carbon and its release from permafrost as it thaws.

The multi-layered methanogenesis scheme improves the representation of high latitude CH_4 emissions, where previous studies underestimated production at cold permafrost sites during "shoulder seasons" (Zona et al., 2016). Figure 2 shows the annual cycle in the observed and modelled wetland CH_4 emissions at the Samoylov Island field site (panel a) and a comparison of observed and modelled annual mean fluxes at this and other sites (panel b). The range of uncertainty used in our study (JULES low Q_{10} - JULES high Q_{10}) captures the range of uncertainty in the observations (In Fig. 2b, the error bars denote the lower and upper estimates from the low and high Q_{10} simulations. The symbols represent the mean value between these estimates). Further, the layered methane scheme used in this work gives a better description of the shoulder season emissions when compared with the original, non-layered methane scheme in JULES. The multi-layered scheme allows an insulated sub-surface layer of active methanogenesis to continue after the surface has frozen. These model developments not only improve the seasonality of the emissions, but more importantly for this study capture the release of carbon as CH_4 from deep soil layers, including thawed permafrost. Further evaluation of the multi-layer scheme can be found in Chadburn et al. (2020).

3. Methane from wetlands: Following Comyn-Platt et al. (2018), we also use the multi-layered soil carbon scheme described in (2) above to give the local land-atmosphere CH_4 flux, E_{CH4} (kg C m^{-2} s^{-1}):

$$E_{CH4} = k \cdot f_{wetl} \cdot \sum_{i=1}^{n\ C_s\ pools} \kappa_i \cdot \sum_{z=0\mathrm{m}}^{z=3\mathrm{m}} e^{-\gamma z} C_{s_{i,z}} \cdot Q_{10}\left(T_{soil_z}\right)^{0.1\left(T_{soil_z}-T_0\right)} \quad (1)$$

where k is a dimensionless scaling constant such that the global annual wetland CH_4 emissions are 180 Tg CH_4 in 2000 (as described in Comyn-Platt et al. (2018)), z is the depth in soil column (in m), i is the soil carbon pool, f_{wetl} (-) is the fraction of wetland area in the grid cell, κ_i (s^{-1}) is the specific respiration rate of each pool (Table 8 of Clark et al. (2011)), C_s (kg m^{-2}) is soil carbon, T_{soil} (K) is the soil temperature. The decay constant γ (= 0.4 m^{-1}) describes the reduced contribution of CH_4 emission at deeper soil layers due to inhibited transport and increased oxidation through overlaying soil layers. This representation of inhibition and of the pathways for CH_4 release to the atmosphere (e.g., by diffusion, ebullition and vascular transport) is a simplification. However, previous work which explicitly represented these processes showed little to no improvement when compared with in-situ observations (McNorton et al., 2016). We do not model CH_4 emissions from freshwater lakes (and oceans).

Comyn-Platt et al. (2018) varied Q_{10} in Eq. 1 to encapsulate a range of methanogenesis process uncertainty. They derive Q_{10} values for each GCM configuration to represent two wetland types identified in Turetsky et al. (2014)

('poor-fen' and 'rich-fen'). They also include a third 'low-Q_{10}', which gives increased importance to high latitude emissions. Their ensemble spread was able to describe the magnitude and distribution of present-day CH_4 emissions from natural wetlands, according to the models used in the then-current global methane assessment (Saunois et al., 2016). Here, we use the 'low-Q_{10}' value of Comyn-Platt et al. (2018) (=2.0) and adopt a 'high-Q_{10}' value of ~4.8 from the rich-fen parameterisation. The two Q_{10} values used here still capture the full range of the methanogenesis process uncertainty.

4. Ozone vegetation damage: We use a JULES configuration including ozone deposition damage to plant stomata, which affects land-atmosphere CO_2 exchange (Sitch et al., 2007). JULES requires surface atmospheric ozone concentrations, O_3 (ppb), for the duration of the simulation period (1850-2100). As in Collins et al. (2018), we do not model tropospheric ozone production from CH_4 explicitly in IMOGEN. Instead, we use two sets of monthly near-surface O_3 concentration fields (January-December) from HADGEM3-A GA4.0 model runs, with the sets corresponding to low (1285 ppbv) and high (2062 ppbv) global mean atmospheric CH_4 concentrations (Stohl et al., 2015). We assume that the atmospheric O_3 concentration in each grid cell responds linearly to the atmospheric CH_4 concentration. We derive separate linear relationships for each month and grid cell, and use these to calculate the surface O_3 concentration from the corresponding global atmospheric CH_4 concentration as it evolves during the IMOGEN run (Sect. 2.2.1). We use CH_4 concentration profiles from the IMAGE SSP2 Baseline and RCP-1.9 scenarios (Sect. 2.3.1), adjusted for natural methane sources (see 3 above and Sect. 2.3.3). We undertake runs using both the 'high' and 'low' vegetation ozone-damage parameter sets (Sitch et al., 2007).

2.2 The IMOGEN intermediate complexity climate model

2.2.1 IMOGEN

The IMOGEN climate impacts model (Huntingford et al., 2010) uses "pattern-scaling" to estimate changes to the seven meteorological variables required to drive JULES. Huntingford et al. (2010) assume that changes in local temperature, precipitation, humidity, wind-speed, surface shortwave and longwave radiation and pressure are linear in global warming. Spatial patterns of each variable (based on the 34 GCM simulations in CMIP5 (Comyn-Platt et al., 2018)) are multiplied by the amount of global warming over land, ΔT_L, to give local monthly predictions of climate change. When using IMOGEN in forward mode, ΔT_L is calculated with an Energy Balance Model (EBM) as a function of the overall changes in radiative forcing, ΔQ (W m^{-2}). ΔQ is the sum of the atmospheric greenhouse gas contributions (Eq. 2) (Etminan et al., 2016), which in the forward mode are either calculated (CO_2 and CH_4) or prescribed (for other atmospheric contributors) on a yearly time step.

$$\Delta Q(total) = \Delta Q(CO_2) + \Delta Q(non\ CO_2\ GHGs) + \Delta Q(aerosols\ and\ other\ climate\ forcers) \quad (2)$$

The EBM includes a simple representation of the ocean uptake of heat and CO_2 and uses a separate set of four parameters for each climate and Earth system model emulated (Huntingford et al., 2010): the climate feedback parameters over land and

ocean, λ_l and λ_o (W m^{-2} K^{-1}) respectively, the oceanic "effective thermal diffusivity", κ (W m^{-1} K^{-1}) representing the ocean thermal inertia and a land-sea temperature contrast parameter, ν, linearly relating warming over land, ΔT_l (K) to warming over ocean, ΔT_o (K), as $\Delta T_l = \nu \Delta T_o$. The climate feedback parameters (λ_l and λ_o) are calibrated using model-specific data for the top of the atmosphere radiative fluxes, the mean land and ocean surface temperatures, along with an estimate of the radiative forcing modelled for the CO_2 changes.

The atmospheric CH_4 concentrations available from the IMAGE database (see Sect. 2.3.1) assume a constant annual wetland CH_4 emission (van Vuuren et al., 2017). However, these emissions have interannual variability and a positive climate feedback (Comyn-Platt et al., 2018;Gedney et al., 2019), and their correct representation is a central part of our study. We follow the same approach that we used in our previous studies (Collins et al., 2018;Comyn-Platt et al., 2018;Gedney et al., 2019). As the IMOGEN-JULES modelling framework does not have an explicit representation of the atmospheric chemistry of methane, we represent the oxidation and hence loss of CH_4 by a single lifetime (τ).

$$\frac{d([CH_4]-[CH_4]_{IMAGE})}{dt} = C\{\sum F[CH_4] - \sum F[CH_4]_{IMAGE}\} - \frac{[CH_4]-[CH_4]_{REF}}{\tau} \quad (3)$$

where $[CH_4]$ and $[CH_4]_{IMAGE}$ are the atmospheric methane concentrations using our new wetland-based, time varying ($F[CH_4]$) and the constant IMAGE ($F[CH_4]_{IMAGE}$) wetland emissions, respectively. Parameter C is a constant to convert from Tg CH_4 to a mixing ratio in parts per billion by volume (ppbv). Further, higher atmospheric concentrations of CH_4 and its oxidation product (carbon monoxide) lower the concentration of hydroxyl radicals, the major removal reaction for CH_4, thereby increasing the atmospheric lifetime of CH_4. Conversely, lower CH_4 concentrations will shorten its atmospheric lifetime. We take account of this feedback of CH_4 on its lifetime (τ), using Eq. 4 (Collins et al., 2018;Comyn-Platt et al., 2018;Gedney et al., 2019), as.

$$\ln(\tau/\tau_o) = s.\ln([CH_4]/[CH_4]_o), \text{ i.e., } \tau = \tau_o \exp(s[CH_4]/[CH_4]_o) \quad (4)$$

In Eq. 4, $[CH_4]_o$ and τ_o are the contemporary atmospheric CH_4 concentration and lifetime, and s is the CH_4-OH feedback factor, defined by $s = \partial \ln(\tau)/\partial \ln(CH_4)$. We take values of τ_o = 8.4 years, $[CH_4]_o$ = 1,745 ppbv and s = 0.28 from Ehhalt et al. (2001) (pages 248 and 250). In our earlier study on the climate-wetland methane feedback Gedney et al. (2019), we investigate the sensitivity to the methane lifetime and the feedback factor, in addition to an analysis of the main drivers on the wetland methane-climate feedback and the main sources of uncertainty. Gedney et al. (2019) conclude that the limited knowledge of contemporary global wetland emissions is a larger source of uncertainty than that from the projected climate spread of the 34 GCMs. We quantify this uncertainty in our experimental design by using two values of Q_{10} (see Sect. 2.1).

In response to our dynamic interactive calculations of atmospheric CH_4 concentrations, we derive the related change in methane radiative forcing (RF). We use the formulation from Etminan et al. (2016), which accounts for the short-wave absorption by CH_4 and the overlap with N_2O. The atmospheric oxidation of methane (by the hydroxyl radical) leads to the production of tropospheric ozone and stratospheric water vapour. We calculate these indirect contributions of methane to the overall radiative forcing, following the approach for methane adopted in our previous work (Collins et al., 2018;Comyn-Platt

et al., 2018;Gedney et al., 2019). Collins et al. (2018) represent the forcing contributions from O_3 and stratospheric water vapour as linear functions of the CH_4 mixing ratio, based on the analysis presented in IPCC AR5 Myhre et al. (2013). The indirect methane forcings amount to 2.36×10^{-4} $\pm1.09\times 10^{-4}$ W m^{-2} per ppb CH_4 (i.e., 0.65±0.3 times the CH_4 radiative efficiency). Hence we incorporate the indirect effects of these CH_4 emission changes by an approximation, multiplying the CH_4 radiative forcing by 1.65.

In this study, we use the inverse version of IMOGEN, which follows prescribed temperature pathways (Fig. 3(a)), to derive the total radiative forcing (ΔQ [total]) and then the CO_2 radiative forcing (ΔQ [CO_2]), using Eq. 2. Comyn-Platt et al. (2018) describe the changes made to the EBM to create the inverse version. As each of the 34 GCMs that IMOGEN emulates has a different set of EBM parameters, each GCM has a different time-evolving radiative forcing (ΔQ) estimate for a given temperature pathway, $\Delta T_G(t)$. When IMOGEN is forced with a historical record of ΔT_G , the range of ΔQ for the near present day (year 2015) from the 34 GCMs is 1.13 W m^{-2}. To ensure a smooth transition to the modelled future, we require the historical period, 1850-2015, to match observations of both ΔT_G and atmospheric composition for all GCMs. As we have a model-specific estimate of the radiative forcing modelled for the CO_2 changes (see above), we, therefore, attribute the spread in ΔQ to the uncertainty in the non-CO_2 radiative forcing component, particularly the atmospheric aerosol contribution, which has an uncertainty range of -0.5 to -4 Wm^{-2} (Stocker et al., 2013). Apart from our modelled CH_4 and CO_2 radiative forcings and the potential future balances between them, we use the projections from the IMAGE SSP2 baseline or RCP1.9 scenario for the radiative forcing of other atmospheric contributors (Fig. 3(b)).

2.2.2 Temperature Profile Formulation

Huntingford et al. (2017) define a framework to create trajectories of global temperature increase, based on two parameters, and which model the efforts of humanity to limit emissions of greenhouse gases and short-lived climate forcers, and, if necessary, capture atmospheric carbon. These profiles have the mathematical form of:

$$\Delta T(t) = \Delta T_0 + \gamma t - (1 - e^{-\mu(t)t})[\gamma t - (\Delta T_{Lim} - \Delta T_0)] \quad (5)$$

where $\Delta T(t)$ is the change in temperature from pre-industrial levels at year t, ΔT_0 is the temperature change at a given initial point (in this case ΔT_0 = 0.89°C for 2015), ΔT_{Lim} is the final prescribed warming limit and

$$\mu(t) = \mu_0 + \mu_1 t \text{ and } \gamma = \beta - \mu_0(\Delta T_{Lim} - \Delta T_0) \quad (6)$$

where β (= 0.00128) is the current rate of warming and μ_0 and μ_1 are tuning parameters which describe anthropogenic attempts to stabilise global temperatures (Huntingford et al., 2017). The parameter values used for the two profiles are: (a) 1.5°C profile: ΔT_{lim} = 1.5°C; μ_0 = 0.1 and μ_1 = 0.0; (b) 2°C profile: ΔT_{lim} = 2°C; μ_0 = 0.08 and μ_1 = 0.0.

2.3 Scenarios and model runs

We undertake a control run and other simulations with anthropogenic CH_4 mitigation or land-based mitigation, stabilising at either 1.5°C or 2.0°C warming without a temperature overshoot. We denote the control run as "CTL", the anthropogenic CH_4 mitigation scenario, a land-based mitigation scenario using BECCS and a variant land-based scenario focussing on AR, as "CH_4", "BECCS", "Natural", respectively. We also undertake runs combining the CH_4 and land-based mitigation scenarios (coupled "BECCS+CH_4" and coupled "Natural+CH_4") to determine if there are any non-linearities when we combine these mitigation scenarios. We summarise the key assumptions of these scenarios in Table 1.

We use future projections of atmospheric CH_4 concentrations and LULUC (specifically, the areas assigned to agriculture and within that to BECCS) from the IMAGE SSP2 projections (Doelman et al., 2018) as input or prescribed data for both the methane and land-based mitigation strategies (Table 1). This ensures that all projections are consistent and based on the same set of IAM model and socio-economic pathway assumptions. The SSP2 socio-economic pathway is described as "middle of the road" (O'Neill et al., 2017), with social, economic, and technological trends largely following historical patterns observed over the past century. Global population growth is moderate and levels off in the second half of the century. The intensity of resource and energy use declines. We define the upper and lower limits of anthropogenic mitigation as the lowest (RCP1.9, denoted "IM-1.9") and highest ("baseline", denoted "IM-BL") total radiative forcing pathways, respectively, within the IMAGE SSP2 ensemble (Riahi et al., 2017). As described in Section 2.2.1, we modify the atmospheric concentrations of CH_4 in the IMOGEN-JULES modelling as the IMAGE scenarios assume constant natural and hence wetland methane emissions.

2.3.1 Methane: baseline and mitigation scenario

The anthropogenic CH_4 emission increase from 318 Tg yr^{-1} in 2005 to 484 Tg yr^{-1} in 2100 in the IMAGE SSP2 baseline scenario, but fall to 162 Tg yr^{-1} in 2100 in the IMAGE SSP2 RCP1.9 scenario. The sectoral CH_4 emissions in 2005 (Energy Supply & Demand: 113; Agriculture: 136; Other Land Use (primarily burning): 18; Waste 52, all in Tg yr^{-1}) are in agreement with the latest estimates of the global methane cycle (Saunois et al., 2019). As summarised in Supplementary Information, Table SI.1, the reduction in CH_4 emissions from specific source sectors is achieved as follows: (a) coal production by maximising CH_4 recovery from underground mining of hard coal; (b) oil/gas production & distribution, through control of fugitive emissions from equipment and pipeline leaks, and from venting during maintenance and repair; (c) enteric fermentation, through change in animal diet and the use of more productive animal types; (d) animal waste by capture and use of the CH_4 emissions in anaerobic digesters; (e) wetland rice production, through changes to the water management regime and to the soils to reduce methanogenesis; (f) landfills by reducing the amount of organic material deposited and by capture of any CH_4 released; (g) sewage and wastewater, through using more wastewater treatment plants and also recovery of the CH_4 from such plants, and through more aerobic wastewater treatment. The levels of reduction vary between sectors, from 50% (agriculture) to 90% (fossil-fuel extraction and delivery). The abatement costs are between US$ 300-1000 (1995 US$) (Supplementary Information, Table SI.1). Figure 4 presents the IMAGE baseline and RCP1.9 CH_4 emission pathways

globally and for selected IMAGE regions, including the major-emitting regions of India, USA and China (Supplementary Information, Figure SI.1 shows the emission pathways for all 26 IMAGE regions). These two methane emission pathways (IMAGE SSP2 baseline and RCP1.9) define our "CTL" and "CH_4" scenarios, respectively.

2.3.2 Land-based mitigation: baseline, BECCS and Natural scenarios

For our land-based mitigation scenarios, we take time series of the annual areas assigned to agriculture (crops and pasture) and within that, the area allocated to bioenergy crops, from the IM-BL and IM-1.9 scenarios (defined at the start of Sect. 2.3). We use the dynamic vegetation module in JULES to calculate the evolution of the natural plant functional types and the non-vegetated surface on the remaining land area in the grid cell (see Land use in Sect. 2.1).

The IM-BL LULUC scenario assumes (a) moderate land-use change regulation; (b) moderately effective land-based mitigation; (c) the current preference for animal products; (d) moderate improvement in livestock efficiencies; and (e) moderate improvement in crop yields (Table 1 in (Doelman et al., 2018)). It represents a control scenario within which agricultural land is accrued to feed growing populations associated with the SSP2 pathway and with no deployment of BECCS. Three types of land-based climate change mitigation are implemented in the IMAGE land use mitigation scenarios (Doelman et al., 2018): (1) bioenergy; (2) reducing emissions from deforestation and degradation (REDD or avoided deforestation); and (3) reforestation of degraded forest areas. For the IM-1.9 scenario, there are high levels of REDD and full reforestation. The scenario assumes a food-first policy (Daioglou et al., 2019) so that bioenergy crops are only implemented on land not required for food production (e.g., abandoned agricultural crop land, most notably, in central Europe, southern China and eastern USA, and on natural grasslands in central Brazil, eastern and southern Africa, and Northern Australia (Doelman et al., 2018)). The IM-1.9 scenario also requires bioenergy crops to replace forests in temperate and boreal regions (notably Canada and Russia). The demand for bioenergy is linked to the carbon price required to reach the mitigation target (Hoogwijk et al., 2009). In this scenario, the area of land used for bioenergy crops expands rapidly from 2030 to 2050, reaching a maximum of 550 Mha in 2060, and then declining to 430 Mha by 2100. Table 2 gives the maximum area of BECCS deployed in each IMAGE region for the IM-1.9 scenario. This defines the land use in the "BECCS" scenario.

We define a third LULUC pathway, which is identical to the "BECCS" scenario, except that any land allocated to bioenergy crops is allocated instead to natural vegetation, i.e., areas of natural land, which are converted to bioenergy crops, remain as natural vegetation, and areas, which are converted from food crops or pasture to bioenergy crops, return to natural vegetation. We make no allowance for any changes in the energy generation system, as this would require energy sector modelling that is beyond the scope of this study. We denote this scenario as "Natural". Table 2 also summarises the main differences in land use between the BECCS and Natural scenarios for each IMAGE region.

Figure 5 presents time series of the land areas calculated for trees and prescribed for agriculture (including bioenergy crops) and bioenergy crops for the "BECCS" and "Natural" scenarios for the Russia and Brazil IMAGE regions, each as a

difference to the baseline scenario (IM-BL). Supplementary Information, Figure SI.2 is equivalent to Fig. 5 for all the IMAGE regions.

2.3.3 Model runs

For each temperature pathway (1.5°C or 2.0°C) and for the baseline and each mitigation scenario, the set of scenario runs comprises a 136-member ensemble (34 GCMs x 2 ozone damage sensitivities x 2 methanogenesis Q_{10} temperature sensitivities). In all model runs, we include the effects of the methane and carbon-climate feedbacks from wetlands and permafrost thaw, which we have shown previously to be significant constraints on the AFFEBs (Comyn-Platt et al., 2018).

As shown in Fig. 1, we use a number of input or prescribed datasets: (a) time series of the annual area of land used for agriculture, including that for BECCS if appropriate; (b) time series of the global annual mean atmospheric concentrations of CH_4 (and N_2O for the radiative forcing calculations of CO_2 and CH_4); (c) time series of the overall radiative forcing by SLCFs and non-CO_2 GHGs (corrected for the radiative forcing of CH_4); and (d) time series of annual anthropogenic CH_4 emissions (used in the post-processing step). We take these from the IMAGE database for the relevant IMAGE SSP2 scenario (baseline or SSP2-1.9). Table 1 lists the main scenario runs, their key features and the prescribed datasets used (for agricultural land and BECCS, anthropogenic emissions and atmospheric concentrations of CH_4 and the non-CO_2 radiative forcing).

Figure 6 presents the effect of these scenarios on the modelled atmospheric CH_4 and CO_2 concentrations. We adjust the input atmospheric CH_4 concentrations to allow for the interannual variability in the wetland CH_4 emissions, as described in Sect. 2.2.1. As we use the same input datasets for the two warming targets, the major control on the modelled atmospheric CH_4 concentrations is the CH_4 emission pathway followed, with the temperature pathway (1.5° versus 2°C warming) having a minor effect. For CO_2, on the other hand, the temperature and the CH_4 emission pathways both lead to increased atmospheric CO_2 concentrations, with the temperature pathway having a slightly larger effect.

2.4 Post-processing

2.4.1 Anthropogenic Fossil Fuel Emission Budget and Mitigation Potential

Following Comyn-Platt et al. (2018), we define the anthropogenic fossil fuel CO_2 emission budget (AFFEB) for scenario *i* as the change in carbon stores from present to the year 2100:

$$AFFEB_i = [C^{land}(2100) - C^{land}(2015)]_i + [C^{ocean}(2100) - C^{ocean}(2015)]_i + [C^{atmos}(2100) - C^{atmos}(2015)]_i + BECCS(2015{:}2100)_i \quad (7)$$

where C^{land} *(t)*, C^{ocean} *(t)* and C^{atmos} (t) are the carbon stored in the land, ocean and atmosphere, respectively, in year *t* and *BECCS*(t_1:t_2) is the carbon sequestered via BECCS between the years t_1 and t_2. The atmospheric carbon store does not include

CH_4. This is a reasonable approximation, however, given the relative magnitudes of the atmospheric concentrations of CH_4 (~2 ppmv at the surface) and CO_2 (400 ppmv).

Within the IMOGEN-JULES modelling framework, we use (a) the IMOGEN climate emulator to derive the changes in the ocean and atmosphere carbon stores, and (b) JULES for the changes in the land carbon store and carbon sequestered through BECCS. We discuss the changes in the carbon stores for the baseline and different mitigation scenarios in Sect. 3.1.

For brevity in the subsequent discussion, we use the following shorthand where the terms on the RHS of Eq. 7 are equivalent to those on the RHS of Eq. 8:

$$AFFEB_i = \Delta C_i^{land} + \Delta C_i^{ocean} + \Delta C_i^{atmos} + BECCS_i \quad (8)$$

We define the mitigation potential (MP) for a mitigation strategy, *j*, as the difference between a control AFFEB ($AFFEB_{ctl}$) and the AFFEB resulting from applying the strategy i.e.:

$$MP_j = AFFEB_j - AFFEB_{ctl} \quad (9)$$

which can be broken down into its component parts as:

$$MP_j = MP_j^{land} + MP_j^{ocean} + MP_j^{atmos}$$

$$MP_j = \left(\Delta C_j^{land} - \Delta C_{ctl}^{land}\right) + \left(\Delta C_j^{ocean} - \Delta C_{ctl}^{ocean}\right) + \left(\Delta C_j^{atmos} - \Delta C_{ctl}^{atmos}\right) + BECCS_j \quad (10)$$

2.4.2 Optimisation of the land-based mitigation

Harper et al. (2018) find that the land-use pathways do not provide a clear choice for the preferred mitigation pathway. The key issue is that replacing natural vegetation with bioenergy crops often results in large emissions of soil carbon and the loss of the benefits of maintaining forest carbon stocks. In such circumstances, Harper et al. (2018) find that the loss of soil carbon in regions with high carbon density makes it difficult for BECCS to deliver a net negative emission of CO_2. Hence, to optimise the land-based mitigation (LBM), we compare the land-carbon stocks in the BECCS and Natural scenarios. We then select the optimum land-management option for each grid cell simulated as that, which maximises the *AFFEB* by year 2100. That is:

$$AFFEB_{LBM} = \Delta C_{BECCS}^{atmos} + \Delta C_{BECCS}^{ocean} + \Delta C_{LBM}^{land} \quad (11)$$

with

$$\Delta C_{LBM}^{land} = \begin{cases} \sum_l^{grid\ cells} \Delta C_{BECCS}^{land} + BECCS & \text{where } \Delta C_{BECCS}^{land} < \Delta C_{BECCS}^{land} + BECCS \\ & or \\ \sum_l^{grid\ cells} \Delta C_{Natural}^{land} & \text{where } \Delta C_{Natural}^{land} > \Delta C_{BECCS}^{land} + BECCS \end{cases} \quad (12)$$

where $\Delta C^{store}_{scenario}$ is the change in carbon between 2015 and 2100 for the 'store' (= atmosphere, ocean or land) for the LULUC scenario. We use the ocean and atmosphere contributions from the BECCS simulations as the changes in store size between the BECCS and Natural simulations are negligible (i.e. <2GtC).

2.4.3 Assumptions about BECCS efficiency

The efficacy of the BECCS scheme implemented in JULES is significantly lower than that of other implementations (Harper et al., 2018), reflecting the importance of assumptions about the efficiency of the BECCS process and bioenergy crop yields in determining their ability to contribute to climate mitigation. More specifically, there is (1) large uncertainty in carbon losses from farm to final storage (Harper et al. (2018) assumed a 40% loss compared to 13-52% loss found in other studies); and (2) a large range in potential productivity of second-generation lignocellulosic bioenergy crops, with JULES falling on the low end. JULES in this study and in Harper et al. (2018) simulated median average yields of ~4.8 and ~4.6 tDM ha^{-1} yr^{-1} , respectively, compared to measured median of 11.5 tDM ha^{-1} yr^{-1} and simulated average of 15.8 tDM ha^{-1} yr^{-1} in IMAGE. The JULES yield of ~4.8 tDM ha^{-1} yr^{-1} corresponds to ~59 EJ yr^{-1} of primary energy, using the maximum area for BECCS from Table 2 of 637.7 Mha and an energy yield of 19.5 GJ t DM^{-1} (Daioglou et al., 2017). Bioenergy supplied 55.6 EJ yr^{-1} or ~10% of primary energy requirement worldwide in 2017 (WBA, 2019). According to Smith et al. (2016), this would increase to ~170 EJ yr^{-1} of primary energy in 2100, for negative emissions of 3.3 Gt Ceq yr^{-1} from BECCS (as required for a 2°C warming target).

As both of these components are assumed to be diagnostics of the simulations, we can modify the contribution of BECCS to the AFFEB via a post-processing scaling factor, κ, which represents the efficiency of (1) and (2) with respect to the JULES parameterisation. That is, Eq. 12 becomes:

$$\Delta C^{land}_{LBM} = \begin{cases} \sum_{l}^{grid\ cells} \Delta C^{land}_{BECCS} + \kappa\, BECCS & \text{where } \Delta C^{land}_{Natural} < \Delta C^{land}_{BECCS} + \kappa\, BECCS \\ & \quad or \\ \sum_{l}^{grid\ cells} \Delta C^{land}_{Natural} & \text{where } \Delta C^{land}_{Natural} > \Delta C^{land}_{BECCS} + \kappa\, BECCS \end{cases} \quad (13)$$

Figure 7 presents maps of the scaling factor required for BECCS to be the preferable mitigation option, as opposed to natural land carbon uptake, for each grid cell for warming of 1.5°C or 2°C. There are large factors in the northern temperate and boreal regions, parts of Africa and Australia. As discussed in Harper et al. (2018), this follows from the loss of soil carbon in the tropics and at high northern latitude leading to long recovery or payback times (10-100+ years and >100 years, respectively, Fig. 6(c) in their paper). The payback time is however insignificant when bioenergy crops replace existing agriculture, for example in Europe and eastern North America.

Additionally, we define a threshold efficiency factor, κ^*, which represents the required BECCS efficiency for BECCS to be a preferable mitigation strategy for a given grid-cell, i.e.:

$$\kappa^* = \frac{\Delta C^{land}_{INatural} - \Delta C^{land}_{BECCS}}{BECCS} \quad (14)$$

This increased efficiency can be considered to be the additional bioenergy harvest (H) and/or the reduced carbon losses from farm to storage needed to pay back the carbon debt accrued due to land-use change (since carbon removed via BECCS = Hε, where ε is the assumed efficiency factor for farm to storage carbon conservation and H is the simulated biomass harvest). In addition, κ^* implies a new threshold (or break-even) level of BECCS:

$$BECCS^* = \kappa^* * BECCS \quad (15)$$

In other words, BECCS* is equivalent to the carbon loss due to the land use change to grow the bioenergy crops. Our IMOGEN-JULES simulations assume a 40% carbon loss from farm to final storage, although other studies have assumed this to be as low as 13% (Harper et al., 2018). To assess the feasibility of meeting this break-even level of BECCS, we calculate the harvest (H^*) that would be needed if carbon losses are to be minimised, i.e. by increasing ε from 0.6 to 0.87, and assuming in Eq. 15 that:

$$BECCS^* = 0.87\ H^* \text{ and } BECCS = 0.60\ H$$

So:

$$H^* = \kappa^* * \frac{0.6}{0.87} * H \quad (16)$$

We discuss this further in Sect. 3.2.

3 Results and Discussion

3.1 Global Perspective

We calculate the anthropogenic fossil fuel emission budget to limit global warming to a particular temperature target as the sum of the changes in the carbon stores of the atmosphere, land (vegetation and soil) and ocean between 2015 and 2100 (Sect. 2.4.1, Eq. 7 and 8). We use a BECCS scale factor (□) of unity. In Fig. 8, we present the median and spread of the AFFEB (as box and whiskers) from the 136-member ensemble, and the individual GCM/ESM contributions to the AFFEBs from the four carbon pools shown (points), for each of the main scenarios modelled using the IMOGEN-JULES or derived in the post-processing optimisation step (see Table 1 for description of the scenarios).

In all the scenarios apart from the BECCS scenario, there is an increase in the land carbon store (shown as positive changes for Coupled (Natural) and Coupled (Optimised) but as smaller negative changes for "CH_4", "Natural" and "Optimised" scenarios. In the "BECCS" scenario, the land carbon change becomes more negative than in the "CTL" scenario, as bioenergy crops replace ecosystems with higher carbon content. In the combined ('coupled') CH_4 and land-based mitigation scenarios, the reduction in the emissions and hence atmospheric concentrations of CH_4 allow increased atmospheric concentrations of CO_2 (Fig. 6). There is increased uptake of carbon by the land, directly because of the increased atmospheric CO_2 concentration

and indirectly through the reduction in O_3 damage. In the coupled "BECCS" scenario, this increased uptake of atmospheric CO_2 is again offset by the land carbon lost through conversion of the land to bioenergy crops. We also find that there is increased uptake of CO_2 by the oceans for all scenarios. A further co-benefit of reducing the CH_4 emissions and allowing more CO_2 emissions is that the oceanic drawdown of CO_2 rises (although it eventually falls to zero under climate stabilisation and there would also be implications for ocean acidification). In Fig. 9(a), we compare the AFFEBs for both the 1.5°C and 2°C temperature pathways. We find that the absolute AFFEBs are 200-300 GtC larger for the 2°C target than the 1.5°C target. These budgets are in agreement with other estimates, which include corrections to the historical period (Millar et al., 2017). In both Figs. 8 and 9, it should be noted that the land carbon store for the "CH_4" mitigation option at -1.4 GtC (median of ensemble) is not visible in these figures. There has however been a net increase in the land carbon store in the "CH_4" scenario when compared to the land carbon store in the control scenario (-70.8 GtC, median of ensemble). This then explains the positive changes shown for the land carbon stores in the coupled "BECCS+ CH_4" and coupled "Natural+ CH_4" scenarios.

Figure 9(b) shows the mitigation potential of each strategy, calculated as the change in the AFFEB from the corresponding control simulation, for the two temperature pathways (Sect. 2.4.1, Eq. 9 and 10). Methane mitigation is a highly effective strategy; the AFFEBs are increased by 188-206 GtC and 193-212 GtC for the 1.5°C and 2°C scenarios, respectively, where the range represents the interquartile range from the 136-member ensemble (34 GCMs x 2 Q_{10} x 2 ozone sensitivities). This AFFEB increase equates to roughly 20-24 years of emissions at current rates for the 1.5°C target. Land-based mitigation strategies also provide significant increases of 51-57 GtC and 56-62 GtC for the 1.5°C and 2°C AFFEB estimates, respectively. This is equivalent to 6-7 years of emissions at current rates. For our BECCS assumptions (see also below), we find that the BECCS contribution is small for the optimised land-based mitigation pathway and that AR are more effective land-based mitigation strategies (Fig. 9(b)). Although the primary challenge remains mitigation of fossil fuel emissions, these results highlight the potential of these mitigation options to make the Paris climate targets more achievable.

Furthermore, the CH_4 and land-based mitigation strategies show little interaction and their potential can be summed to give a comparable result to the coupled simulation (coupled vs linear in Fig. 9(a) and (b)). This decoupling is despite the CH_4 emissions from the agricultural sector being influenced by land use choices. We can effectively treat the two mitigation strategies as independent, and their sum approximates the combined potential. Such linearity enables simpler and more direct comparisons.

Despite the substantial differences in the absolute AFFEBs for the 1.5° and 2°C targets, the mitigation potential of the CH_4 and land-based strategies is similar for the two temperature pathways considered. This similarity suggests that the mitigation strategies are robust to the target temperature; whether the international community aims for the 1.5° or 2°C target, afforestation, reforestation, reduced deforestation and CH_4 mitigation are beneficial mitigation approaches.

For both temperature pathways (i.e., 1.5°C or 2°C of warming), we investigate the contribution to the uncertainty range from 'climate' as represented by the 34 GCMs emulated and from the land processes investigated (Sect. 2.1). A GCM with higher climate sensitivity will have a lower AFFEB for a specific warming target (and vice versa). In our post-processing steps, we derive a number of statistical parameters from the complete 136-member or the 34-member GCM ensemble for the

individual factorial runs (low Q_{10}/low O_3, low Q_{10}/high O_3, high Q_{10}/low O_3 and high Q_{10}/high O_3), such as mean, standard deviation, median, and various percentiles. Our focus is on the contribution different factors make to the overall standard deviation of the 136-member ensemble (σ_{All}). By factoring out the climate variation (via their means), we calculate the standard deviation for the land processes investigated (σ_{land}). With a knowledge of the overall standard deviation and that for land-only processes, we derive the contribution from 'climate' ($\sigma_{climate}$) assuming that the variance are independent and can be summed (Eq. 17). The contributions of uncertainty are by comparing ratios of σ_{land} to $\sigma_{climate}$.

$$\sigma_{all}^2 = \sigma_{climate}^2 + \sigma_{land}^2 \quad (17)$$

We present the results of this analysis in Table 3 for the Anthropogenic Fossil Field CO_2 Emission Budgets and the Mitigation Potential (= scenario – "CTL") for the 1.5°C temperature profile (Supplementary Information, Table SI.2 is equivalent table for the 2°C temperature profile). Our overall finding is that the climate uncertainty dominates the uncertainty of the AFFEBs. However, when considering different trade-offs between land uncertainty and mitigation options, the impact of climate uncertainty is much weaker. Within the land uncertainty, the O_3 vegetation damage appears to make the greater contribution (from the changes in the mean). Although there is some variation in the ratio ($\sigma_{climate}$:σ_{land}) between the scenarios (0.32±0.13, mean ± standard deviation), this gives us confidence in the robustness of the uncertainty estimates derived, across the scenarios and the 2 temperature profiles.

3.2 Sensitivity to BECCS Efficiency

The BECCS parameterisation used here makes BECCS less effective compared to those in other studies (van Vuuren et al., 2018). Globally across the two temperature targets, our simulations imply a removal of 27-30 GtC from the active carbon cycle via BECCS in the original "BECCS" scenario run, which is reduced to ~7-12 GtC after we optimise the land-use scenario. These removal rates are significantly lower than other estimates based on the same land-use scenarios: 73 GtC in a similar dynamic global vegetation model (LPJ-GUESS) and 130 GtC in IMAGE (Harper et al., 2018). We find that doubling the carbon captured with BECCS in our simulations (Sect. 2.4.3, κ=2) has a relatively small impact on the total mitigation potential in the optimised scenario (Fig. 10(a)). This low sensitivity is because the increased carbon removed by BECCS often accompanies a comparable decrease in the carbon uptake from the "natural" vegetation that it replaces. It is only when setting the BECCS carbon sequestration at 3-5 times its original value that there is a notable increase of the global AFFEB. Further, as shown in Fig. 10(b), there is reduction in soil carbon in specific regions (e.g. Northern temperate and boreal regions), which makes BECCS less effective for carbon sequestration than natural land management options (or there is a long payback time as discussed in Harper et al. (2018)).

Increased carbon removal with BECCS could be realised through either (1) minimizing the loss of carbon from farm to final storage (ε in Sect. 2.4.3), or (2) maximizing the productivity of the bioenergy crop. Our IMOGEN-JULES simulations assume a 40% carbon loss from farm to final storage, although other studies have assumed this to be as low as 13% (Harper et

al., 2018). The bioenergy crop yields in JULES (Fig. 10(c)) are lower than the median yield of Miscanthus (11.5 tons of dry matter (ton DM) ha^{-1} yr^{-1}), measured from 990 mostly European plots (Li et al., 2018), and are about half the productivity of those in the IMAGE simulations. We calculate for each IMOGEN grid cell the increase in carbon removed via BECCS and the associated increase in bioenergy crop yields (H^* in Sect. 2.4.3) required for BECCS to be the preferred mitigation option (Fig. 10(d)), rather than natural land carbon uptake, and assuming minimal amounts of carbon are lost during the BECCS lifecycle (13% carbon loss). In many places, we find that the required yield increases from <10 to 10-20 ton DM ha^{-1} yr^{-1} are achievable, but required yields of > 30 ton DM ha^{-1} yr^{-1} would be more difficult to realise, given the range of yields observed (Li et al., 2018). We provide additional information in the Supplementary Information, Tables SI.4a-SI.4d on the modelled bioenergy yields and the yields required for bioenergy crops to be the preferred land-based mitigation option by IMAGE region. The tables also show that area of bioenergy crops and carbon sequestered by BECCS increases, as expected, with the BECCS scale factor (κ),

We conclude that our uncorrected simulations are a lower estimate for the potential of carbon removal via BECCS. We provide a more optimistic estimate of the BECCS potential using $\kappa = 3$, which results from doubling the JULES yields and increasing the efficiency ε from 0.6 to 0.87 (i.e., $\kappa \sim 2 \times 0.87 / 0.6$). We now find the global land-based mitigation potential to be 88-100 GtC across the two temperature targets, as shown in Fig. 9(c) and (d). Supporting Information, Figure SI.3 shows the corresponding plots for the 2°C warming target. We use $\kappa = 3$ in the subsequent analysis of regional mitigation options and of BECCS water requirements.

3.3 Regional Analysis

We consider the sub-continental implications of CH_4 and land-based mitigation options, using the 26 regions of the IMAGE model (Stehfest et al., 2014). Figure 11 shows the contributions of the three mitigation options - CH_4, carbon uptake through AR and BECCS - to the AFFEBs for each IMAGE region and for the temperature pathway stabilising at 1.5°C.

We estimate the regional land-based mitigation as the change in the land-carbon stores plus the carbon removal via BECCS for each IMAGE region in the IMOGEN-JULES model output. In this accounting, the region where the bioenergy crops are grown is credited with the carbon removal via BECCS. We assume a three-fold increase in carbon removal via BECCS compared to our default simulations (κ=3) to highlight regions where BECCS is potentially viable. Figure 12 shows the sensitivity of the global AFFEBs and Mitigation Potential for κ = 1, 2 and 3 for 1.5□C of warming (Supplementary Information, Figure SI.3 is the corresponding figure for 2□C of warming). For CH_4, we use regional scale factors to allocate changes in the global atmospheric CH_4 concentration, and therefore the CH_4 mitigation potential, to each region, as shown in Supplementary Information, Table SI.3. To derive the regional scale factors, we separately sum the projected anthropogenic CH_4 emissions between 2020 and 2100 between the IMAGE SSP2-Baseline and SSP2-1.9 scenarios (van Vuuren et al., 2017). We calculate the scale factor as the regional fraction of the global difference in the summed emissions (Supplementary Information, Table SI.3). These two CH_4 scenarios are consistent with the CH_4 concentration pathways considered in the CH_4

scenario simulations (Sect. 2.3). We use the scale factors to produce Fig. 11 and 12 (and Supplementary Information, Figures SI.3 and SI.4).

CH_4 mitigation is an effective mitigation strategy for all regions, and especially the major methane emitting regions: India, S. Africa, USA, China and Australasia. Figure 4 presented time series of the anthropogenic CH_4 emissions for selected IMAGE region from 2000 to 2100 (and Supplementary Information, Figure SI.1 presents emission time series for all IMAGE regions). The mitigation of CH_4 emissions from fossil-fuel production, distribution and use for energy is the largest contributor for India, S. Africa, USA, China and Australasia. The emissions from agriculture-cattle (for India, USA and China) and rice production (China and other Asian regions) make smaller contributions.

The impact of the land-based mitigation options links strongly to the managed land-use and land-use change (LULUC). As discussed in Sect. 2.3.2, we list in Table 2 the maximum area of BECCS deployed in each IMAGE region and the main differences in land use between the BECCS and Natural scenarios. Figure 5 presents time series of the land areas calculated for trees and prescribed for agriculture (including bioenergy crops) and bioenergy crops for the BECCS and Natural scenarios for the Russia and Brazil IMAGE regions, each as a difference to the baseline scenario (IM-BL) (see Supplementary Information, Figure SI.3 for all the IMAGE regions). The West Africa region shows the largest natural land carbon uptake (WAF in Fig. 12). Here, there is conversion of crop and pasture to forest, with little land used for bioenergy crops for BECCS. For Brazil (Fig. 5(a)) and the rest of South America, both bioenergy crops and forest expand at the expense of agricultural land. For many other regions, notably Canada, Russia, W. & C. Europe, China, Oceania, there is less carbon uptake from the 'land' in the optimised mitigation scenario, even though the overall carbon uptake has increased. For Canada and Russia, this results from the loss of forest in the BECCS land use scenario (see Fig. 5(b) and Supplementary Information, Figure SI.3). The carbon uptake by BECCS increases as □ increases from 1 to 3 because there are more grid cells where 'BECCS' is the preferred mitigation option in the optimisation process, as evidenced by the increase in area of bioenergy crops (Supplementary Information, Tables SI.3a and SI.3c). As □ only affects the 'BECCS' term (Sect. 2.4.3, Eq. 13), the increased carbon removed by BECCS is often accompanied by a decrease in the carbon uptake from the "natural" vegetation that it replaces. This can be seen more clearly in Fig. 12 (and Supplementary Information, Figure SI.3 for 2°C warming) and the Supplementary Information, Tables SI.3b and SI.3d . The version of JULES used in this study currently lacks a fire regime. There will be risks to long-term storage of carbon stored in vegetation in regions with significant areas of fire-dominated vegetation cover (e.g. savannah in Brazil and Africa). Further, this version of JULES does not include a nitrogen cycle, which has been implemented in more recent versions of the model. This will enable the impact of changes in land use and agriculture on N_2O emissions to be integrated into the assessments.

There is relatively little difference in the additional allowable carbon emission budgets introduced by CH_4 and/or the land-based mitigation between 2015 and 2100 for the two temperature pathways considered (Supplementary Information, Figure SI.4 for the contributions at 2□C of warming).

3.4 Water Resources

Smith et al. (2016) estimate the global water requirements for different negative emission technologies, including BECCS. We also derive the water requirements from the carbon uptake by BECCS for our optimised land-based mitigation scenarios. The IM-1.9 land use scenario (Sect. 2.3.2) assumes that bioenergy crops are grown sustainably and are rain-fed (Hoogwijk et al., 2005;Daioglou et al., 2019). Our land surface modelling system explicitly accounts for this. We derive the additional water requirements for BECCS, using $\kappa = 3$ and assuming (a) a marginal increase in water use of 80 m^3 $(tC\ eq)^{-1}$ yr^{-1} when replacing the average short vegetation (i.e., C3/C4 grasses in JULES) by a biomass energy crop (Smith et al., 2016); and (b) 450 m^3 $(tC\ eq)^{-1}$ yr^{-1} for the CCS component (Smith et al., 2016).

Following Postel et al. (1996), we derive the accessible runoff, using their assumptions that only 5% of the total runoff is geographically and/or temporally accessible for the Brazil, Russia and Canada IMAGE regions, and 40% elsewhere. Our present-day estimates of the global annual runoff (43,000-44,200 km^3 yr^{-1}) and the accessible runoff for human use (11,400-11,720 km^3 yr^{-1}) (see Fig. 13) are both in agreement with the values given in Postel et al. (1996), i.e., total and accessible runoffs of 40,700 and 12,500 km^3 yr^{-1}, respectively.

We use the water withdrawals for each IMAGE region given in the IMAGE-SSP2-RCP2.6 scenario for the water demand for agricultural irrigation (Rost et al., 2008) and for other human activities, such as energy generation, industry and domestic usage (Bijl et al., 2016), between 2015 and 2100 (Table 4a and 4b). We assume the same water demands from these sectors for both the 1.5°and 2°C warming targets.

Figure 14 compares the accessible water with the water demand for BECCS and other human activities for the regions that produce a substantial amount of BECCS: Canada, USA, Brazil, Europe, Russia, China, Southern Africa and Oceania for the optimised land-based mitigation. Table 4a and b show the additional water requirements of BECCS calculated for 2060 and 2100, respectively, for the 2°C warming target. We find that the additional demand for BECCS would lead to an exceedence (or use >90%) of the available water for the Oceania and Rest of Southern Africa regions. We also find that the additional demand for BECCS is greater than the total water withdrawals from anthropogenic activities for the Canada and Brazil IMAGE regions. Our estimates represent a maximum possible water usage for BECCS as (i) the SSP2 scenario used already accounts for the lower power generation efficiencies and hence higher water requirements in switching from fossil fuels to bioenergy crops (which could be up to 20-25%) and (ii) the figure used for the CCS component does not allow for future technological improvements in water use. For example, Fajardy and Mac Dowell (2017) indicate a 30-fold reduction in water use when changing from a once-through to a recirculating cooling tower. Our results are less severe than other studies considering BECCS water requirements (Séférian et al., 2018;Yamagata et al., 2018), because the carbon removed by BECCS in this study (30 GtC) is already limited to regions where it is more beneficial to the AFFEB than forest-based mitigation options. We also note from Bijl et al. (2016) that the water demand for irrigation, derived using the coupled IMAGE-LPJmL models, is low compared to other estimates in the literature. Higher water demand for irrigation existing agriculture would be an additional constraint on the water available for BECCS. Nevertheless, our results indicate that the additional water demand

for BECCS would have large impacts in half of the regions substantially invested in BECCS: Oceania, Rest of South Africa, Brazil and Canada.

4 Conclusions

Our paper brings together previous studies that looked separately into the potential of methane mitigation (Collins et al., 2018) and land-management options (especially forest conservation and BECCS) (Harper et al., 2018), into a single unified framework. Uniquely, this allows us to compare these options at local and regional scales. We utilise the detailed JULES land-surface model, which includes methane production from wetlands and permafrost thaw (Comyn-Platt et al., 2018) and the effect of CH_4 emissions on land carbon storage via ozone impacts on vegetation (Sitch et al., 2007), and also span the range of climate model projections using the IMOGEN ESM-emulator. For each temperature pathway and each of the three mitigation options, the set of scenario runs comprises a 136-member ensemble (34 GCMs x 2 ozone damage sensitivities x 2 methanogenesis Q_{10} temperature sensitivities).

This analysis quantifies the regional differences in potential CH_4 and/or land-based strategies to aid mitigation of climate change. We present our findings within a full probabilistic framework, capturing uncertainty in climate projections across the CMIP5 ensemble, as well as process uncertainties associated with the strength of natural CH_4 climate feedbacks from wetlands and ozone-induced vegetation damage. Globally, mitigation of anthropogenic CH_4 emissions and the optimised land-based mitigation can potentially offset (i.e. allow extra) fossil fuel carbon dioxide emissions of 188-212 GtC and 51-100 GtC, respectively. These bounds are almost independent of the eventual global-warming target, or the climate sensitivity of the climate models emulated. As shown in Sect. 3.1, the CH_4 and land-based mitigation strategies show little interaction and their potential can be summed to give a comparable result to the corresponding coupled simulation. This decoupling is despite the CH_4 emissions from the agricultural sector being influenced by land use choices. We can therefore treat the two mitigation strategies as independent, and sum their individual potentials. Such linearity enables simpler and more direct comparisons between the carbon budgets of methane and land-based mitigation strategies. Some caveats remain however. Land surface models still require refinement, alongside improved characterisation of the assumptions inherent in the socio-economic pathways and IAM modelling. Further, we do not allow for the reduced emissions from fossil fuel combustion due to the bioenergy crop being grown (or the converse when bioenergy crops are replaced in the Natural model run), as this would require energy sector modelling that is beyond the scope of this study.

For the "Natural" land-based scenario (see Table 1), we find a mitigation potential of 50-55 GtC (183-201 $GtCO_2$). The land-based mitigation estimates vary over wide ranges, partly related to different assumptions on land use and carbon pools. Our results are within the wide range of the overall deployment of CO_2 removal by Agriculture, Forestry and Other Land Use (including afforestation and reforestation) to 2100 of 200 [0-550] $GtCO_2$ (Page 2.40 in IPCC (2018)) and of estimates of the cumulative potential to 2100 from 80 to 260 $GtCO_2$ (Table 2) in Minx et al. (2018). In the "BECCS" scenario, we obtain a geological carbon storage via BECCS (27±1 GtC median, interquartile range) similar to that (30±1 GtC) derived by Harper et

al. (2018), for the same land use scenario (IM-1.9). Our result is lower as we include the natural methane feedbacks from wetlands and permafrost thaw. Inclusion of this better process description leads to ~10% reduction in carbon budgets(Comyn-Platt et al., 2018). These estimates for the geological carbon storage via BECCS are much lower than the corresponding value derived by the IMAGE IAM (130 GtC). Harper et al. (2018) discuss this difference, identifying a number of reasons for the lower value: the use of initial above ground biomass harvested in boreal forests for BECCS, the replacement of fossil-fuel based emissions in the energy system, as well as specific assumptions about crop yields, conversion efficiency, use of residues, the proportion of bioenergy crops used with CCS. Estimates of the BECCS contribution in the literature vary over a wide range (from 178 to >1000 $GtCO_2$, according to Minx et al. (2018)), but in recent studies these result are typically revised downwards taking into account among others sustainability constraints (e.g. Fuss et al. (2018) suggests a potential of 0.5-5 $GtCO_2$ per year in 2050).

We investigate the efficacy of our "BECCS" scenario by increasing the productivity of BECCS (using a scale factor □). From comparison with observed bioenergy crop yields, we argue that the scale factor could be between 1 and 3. We highlight how using this range of □ provides characterisation of an additional source of uncertainty on the land-based mitigation potential. In our optimised land-based mitigation scenario, which maximises the land carbon uptake (Sect 2.4.2, Eq. 10), the increased carbon removed by BECCS is often accompanied by a decrease in the carbon uptake from the "natural" vegetation that it replaces (as discussed in Sect. 3.3 and shown in Figure 12). This concern is equivalent to the statement in Harper et al. (2018) that the "use of BECCS in regions where bioenergy crops replace ecosystems with high carbon contents could easily result in negative carbon balance". Hence the particularly novel feature of our paper is that our optimal approach accounts explicitly for that trade-off, only suggesting BECCS where there is a net gain. For boreal forest regions there is a preference for avoided deforestation, whereas in tropical forest regions both AR and avoided deforestation offer significant potential. From a carbon sequestration perspective, growing bioenergy crops for BECCS is only preferable where it replaces existing agricultural land. BECCS has particular potential if productivities and power production efficiencies are towards the upper limit of expected photosynthetic capability, whilst noting the strong water demand of such crops requires consideration in the context of a growing population.

Stabilising the climate primarily requires urgent action to mitigate CO_2 emissions. However, we estimate that CH_4 mitigation may offset up to 188-212 GtC of anthropogenic CO_2 emissions, while still meeting the same global-warming targets. This offset is a direct consequence of the reduced radiative forcing by methane and of carbon cycle gains. These balances and related flexibilities have the potential to make the Paris targets more achievable. Our range of additional CO_2 emissions broadly applies to both the 1.5° and 2°C warming targets, as the mitigation potential of the CH_4 scenario is similar for the two temperature pathways considered. Although there are differences in the precise methane emission scenarios used, our mitigation potential is similar to that given in Collins et al. (2018). That paper presents values of 155 or 235 GtC for offsetting CH_4 mitigation from a high to a medium or from a high to a low emission scenario, respectively. Our value, and those of Collins et al. (2018), can be compared to the increase of 130 GtC in the carbon budget between a no and a stringent CH_4 emission mitigation scenario estimated by Rogelj et al. (2015). More recently, Harmsen et al. (2020) have also investigated

the mitigation potential of methane, although their results are expressed in terms of changes in radiative forcing and temperature, rather than carbon budgets. An advantage of our analysis remains the inclusion of climate response to altered radiative forcing, enabling understanding in terms of actual CO_2 emissions. We conclude that CH_4 mitigation would be effective globally as a contribution to constraining global warming, and especially so for the major CH_4-emitting regions of India, USA and China.

Code and Data Availability

The IMOGEN-JULES source code used in this work is available from the JULES code repository (https://code.metoffice.gov.uk/trac/jules/browser/main/branches/dev/annaharper/r7971_vn4.8_1P5_DEGREES_CCS, at JULES revision 14477, user account required). The rose suites used for the specific IMOGEN-JULES runs are: u-as624, u-at010, u-at011, u-at013, u-av005, u-av007, u-av008, u-av009, u-ax327, u-ax332, u-ax455, u-ax456, u-ax521, u-ax523, u-ax524, u-ax525, u-bh009, u-bh023, u-bh046, u-bh081, u-bh084, u-bh098, u-bh103 and u-bh105. These can be found at https://code.metoffice.gov.uk/trac/roses-u/ (user account required).

The IMOGEN-JULES source code is also available as a zipped tarball from https://github.com/GarryHayman/Regional_Mitigation_Paper, as are the python scripts used for post-processing. Relevant outputs from the IMOGEN-JULES runs will be made available through a publically accessible data repository (tbc).

Requests for information about the code, data and parameterisations can be made to the corresponding author.

Author Contributions

G.H., C.H., E.C-P., A.H., P.C., T.P., J.H., W.C., J.L. and S.C. designed the IMOGEN runs. All authors contributed to the interpretation of the results and to the writing of or review of the paper. C.H. provided IMOGEN parameters calibrated against the CMIP5 database, and E.C-P and C.H. led the development of the inverse IMOGEN model version. The following specific contributions were also made: (a) E.B., S.C. and N.G.: code and expertise on permafrost, soil carbon and wetland methane modelling, respectively; (b) A.H. and T.P.: land use change data; (c) W.C. and C.W.: ozone ancillary data; (d) D.P.vV. and J.C.D.: IMAGE scenario data on land use, anthropogenic methane emissions and water consumption and withdrawals, and (e) S.S.: expertise on the ozone damage effects.

Competing interests

The authors declare no competing interests.

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Figures

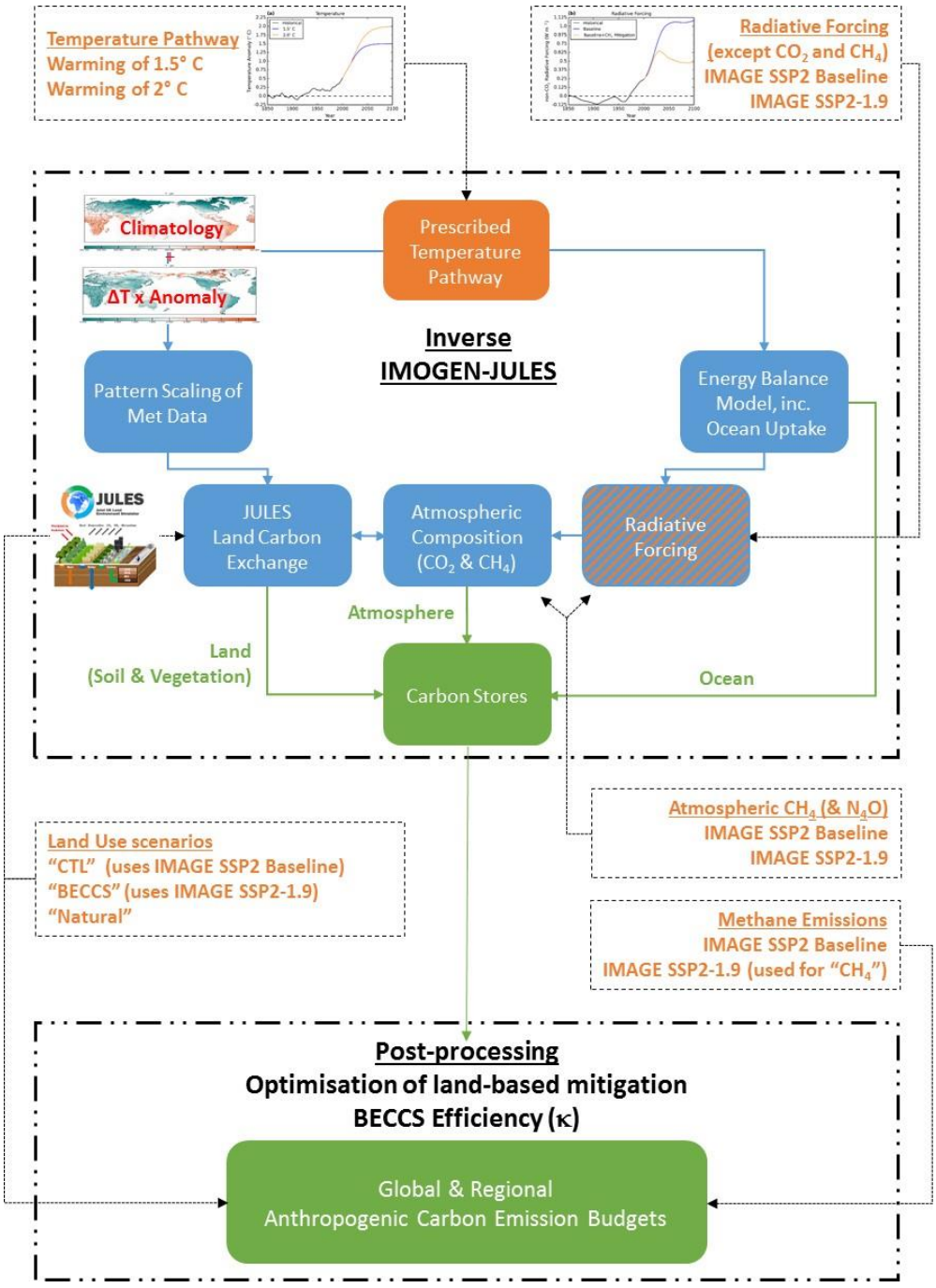
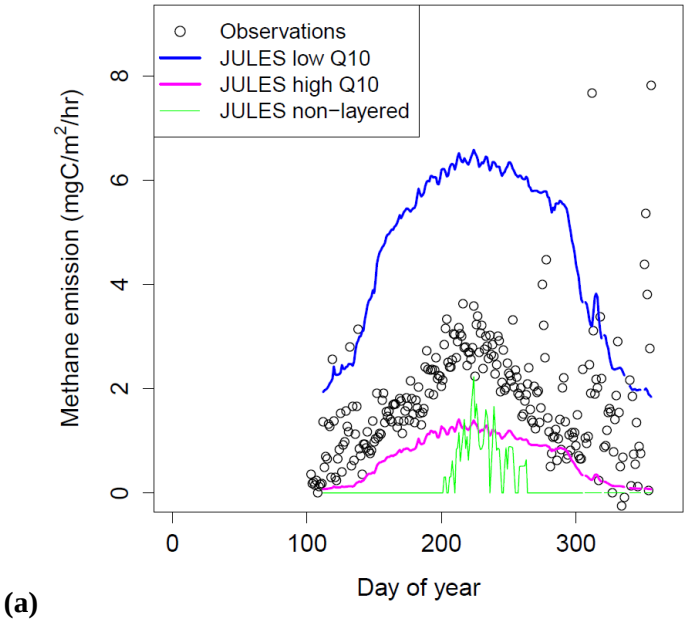
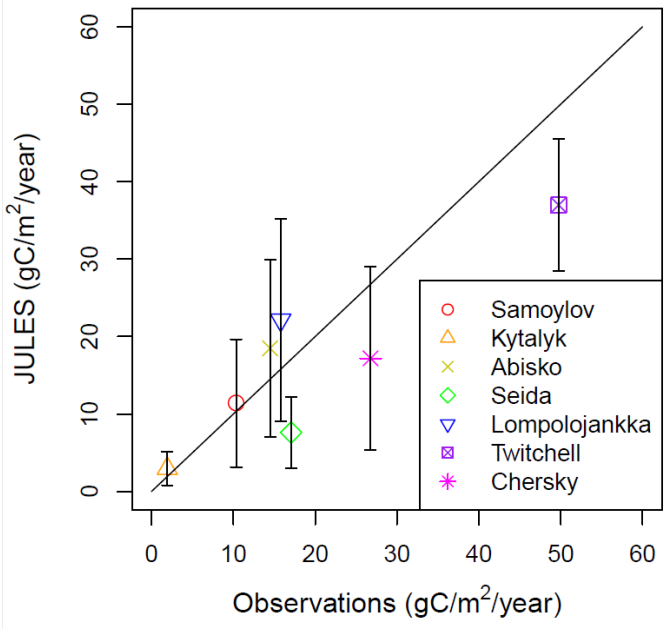


Figure 1 | Schematic of the modelling approach and the workflow. The coloured boxes and text show (a) the key components of the inverted IMOGEN-JULES model (blue), the prescribed and input data used in this study (orange) and the outputs (green).



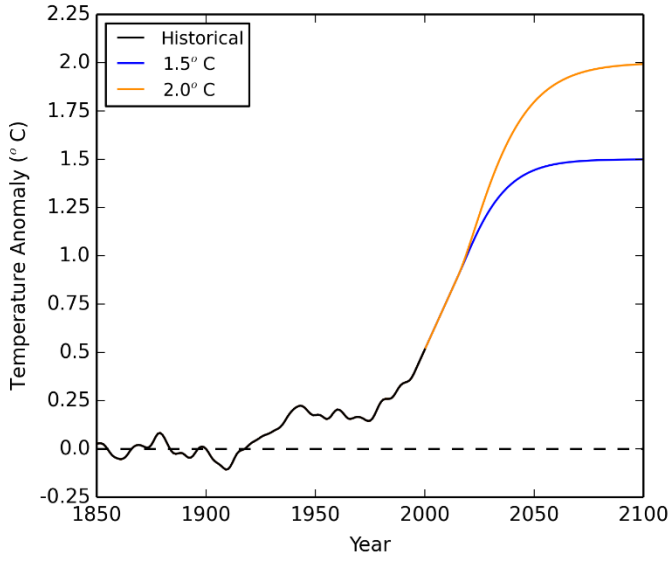
(a)



(b)

Figure 2 | (a) Observed (circles) and modelled wetland methane emissions at the Samoylov Island field site. Modelled wetland methane emissions are shown for the standard JULES non-layered soil carbon configuration (green) and for the JULES layered soil carbon configurations with the low (blue line) and high (magenta line) Q_{10} temperature sensitivities; the low Q_{10} configuration gives higher methane emissions at high-latitude sites such as the Samoylov Island field site. The methane emission data is preliminary and was provided by Lars Kutzbach and David Holl. (b) Comparison of observed and modelled annual mean wetland CH_4 emission fluxes at a number of northern high-latitude and temperate sites. The error bars denote the lower and upper estimates from the low and high Q_{10} simulations. The symbols represent the mean value between these estimates.

(a) Time series of the prescribed temperature pathways



(b) Input time series of the input non-CO_2 radiative forcing

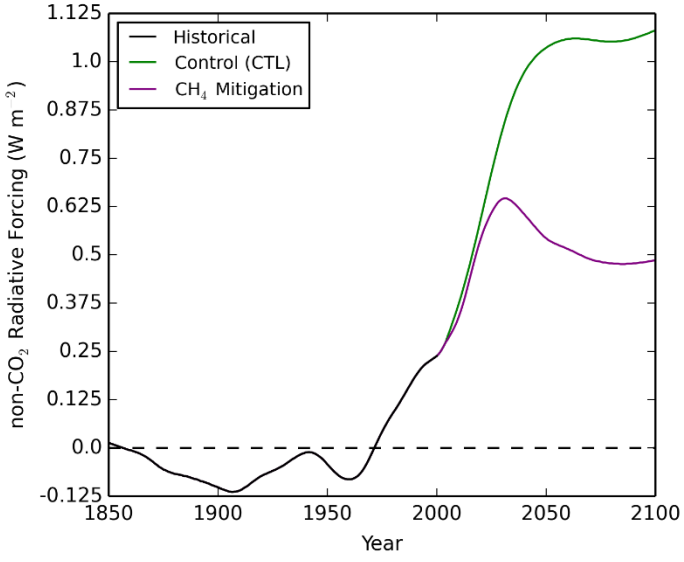


Figure 3 | Time series of key datasets used in the study: (a) the historic temperature record (black) and the prescribed temperature profiles used to represent warming of 1.5°C (blue) and 2°C (orange); (b) the historic (black) and the projected non-CO_2 greenhouse gas radiative forcing (W m^{-2}) for the control (greem) and methane mitigation (purple) scenarios.

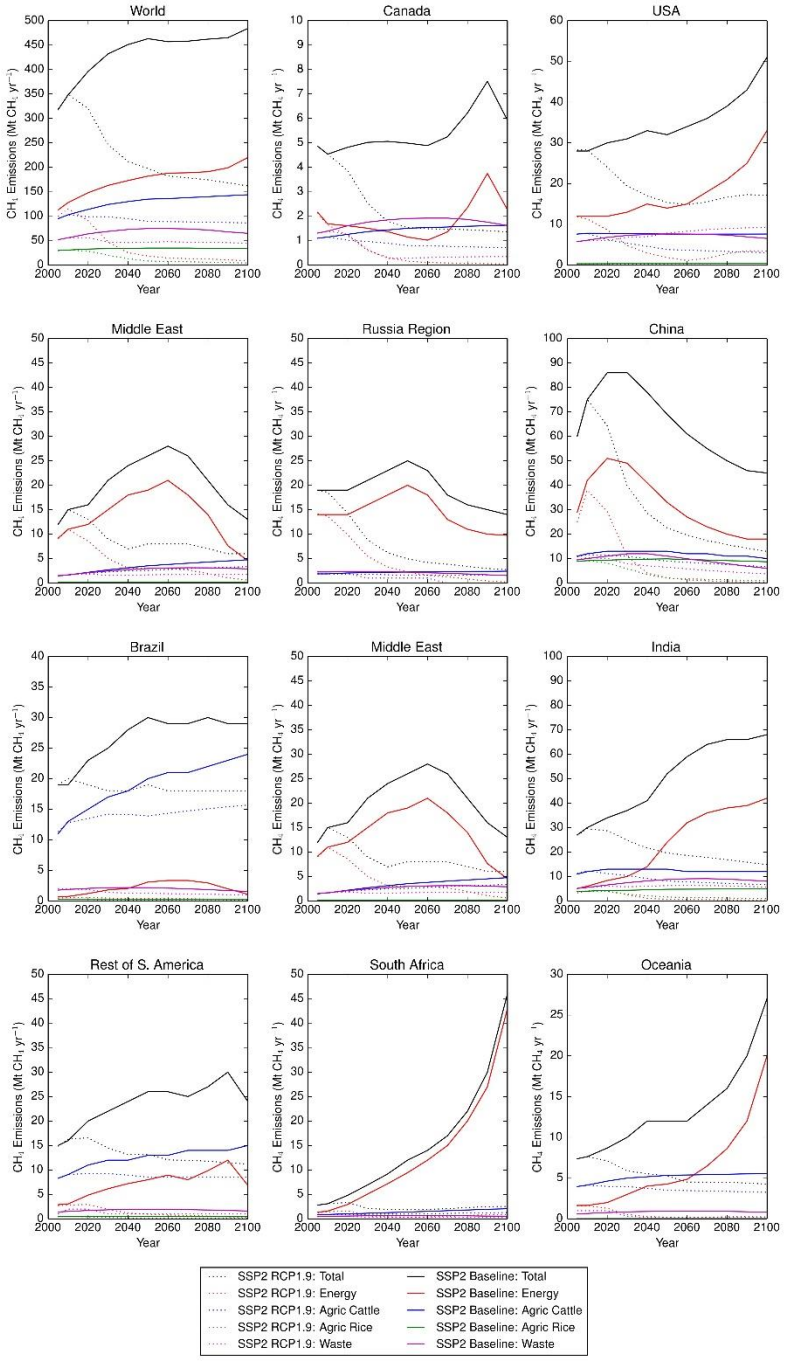


Figure 4 | Time series of annual methane emissions between 2005 and 2100 from all and selected anthropogenic sources according to the IMAGE SSP2 Baseline (solid lines) and SSP2-RCP1.9 (dotted lines) scenarios, globally and for selected IMAGE regions, with total emissions in black, energy sector in red, agriculture-cattle in blue, agriculture-rice in green and waste in magenta. Note the y-axes have different scales for clarity.

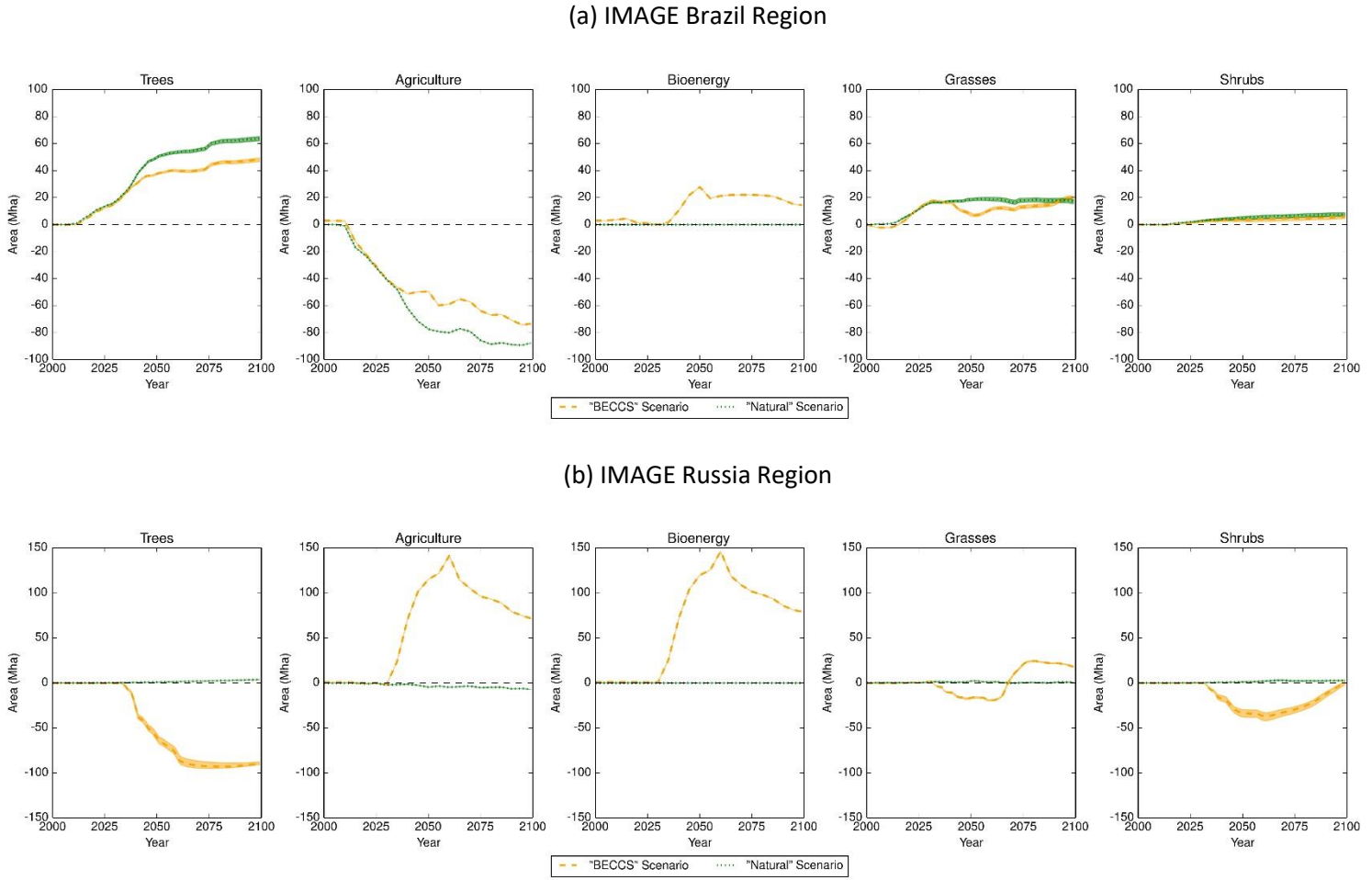


Figure 5 | Time series of the land areas (in Mha) calculated for trees and prescribed for agriculture (including bioenergy crops) and bioenergy crops for the 'BECCS' (orange) and 'Natural, (green), as a difference to the baseline scenario (IM-BL), for Brazil (panel a) and the Russia (panel b) IMAGE regions between 2000 and 2100. The dotted lines are the median and the spread the interquartile range for the 34 GCMs emulated and 4 factorial sensitivity simulations.

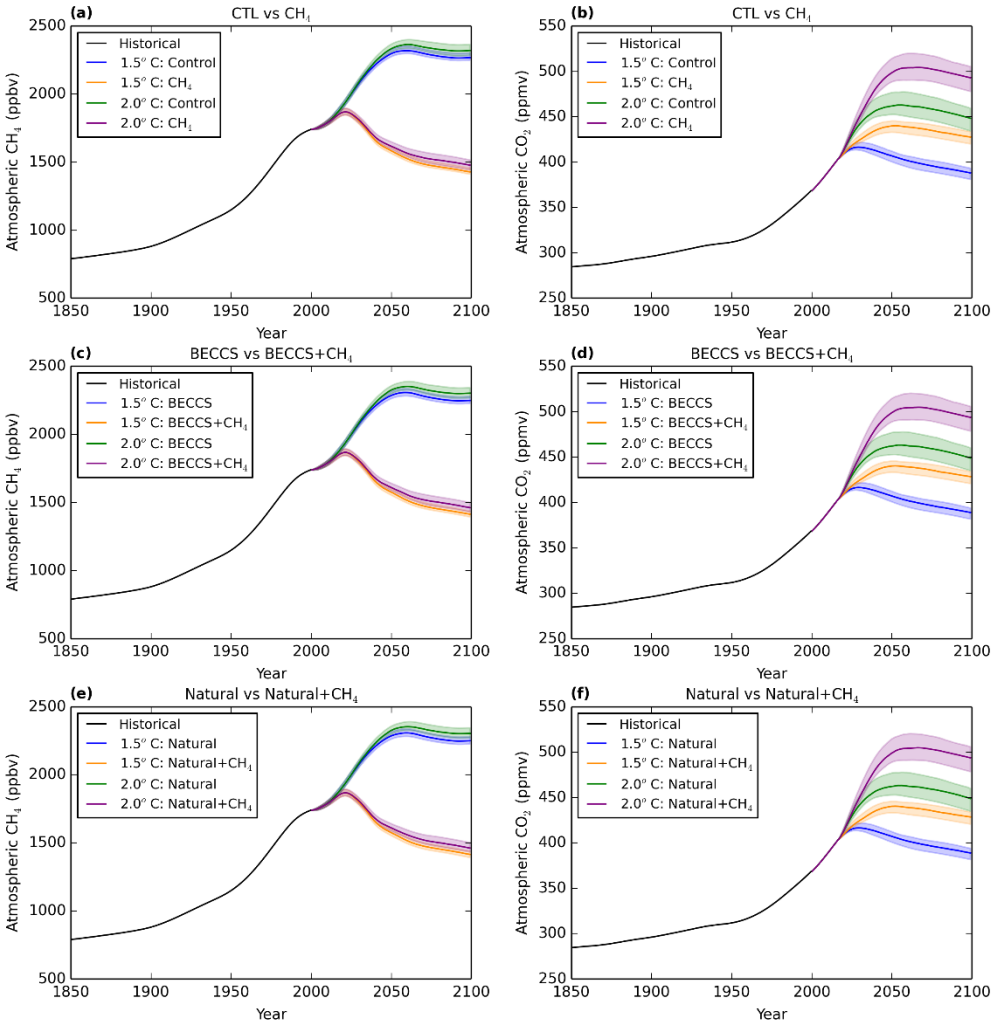
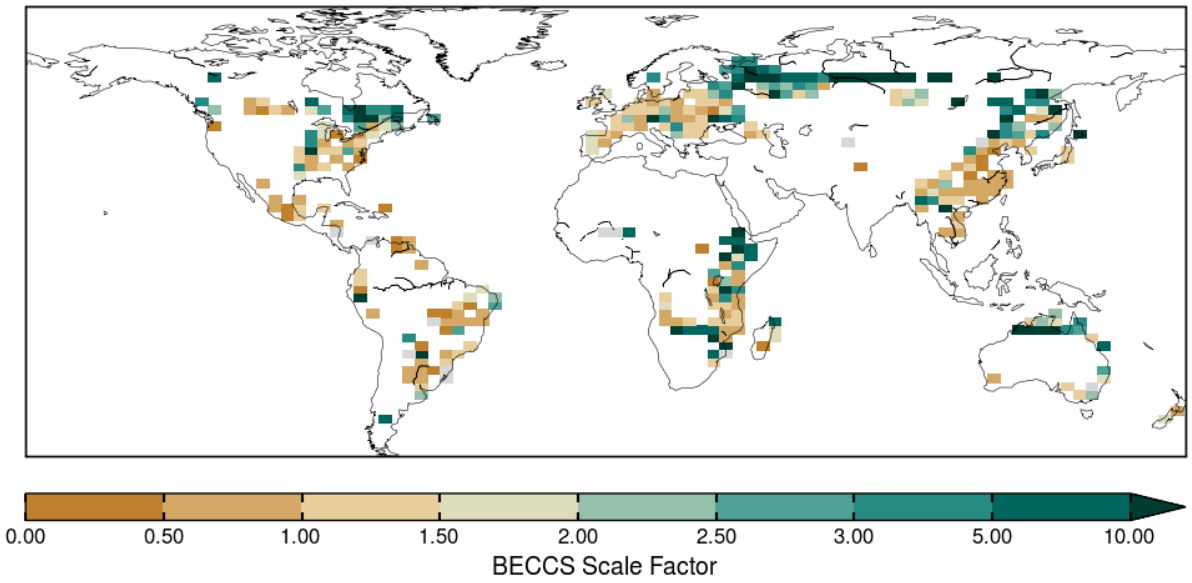


Figure 6 | (a, c, e) Time series of the ensemble median atmospheric CH_4 concentrations (with interquartile range as spread) derived for each temperature profile for the scenarios: (a) "CTL" and "CH_4", (c) "BECCS" and "BECCS+CH_4", (e) "Natural" and "Natural+ CH_4". (d, f, h) show the corresponding time series for the atmospheric CO_2 concentrations.

(a)



(b)

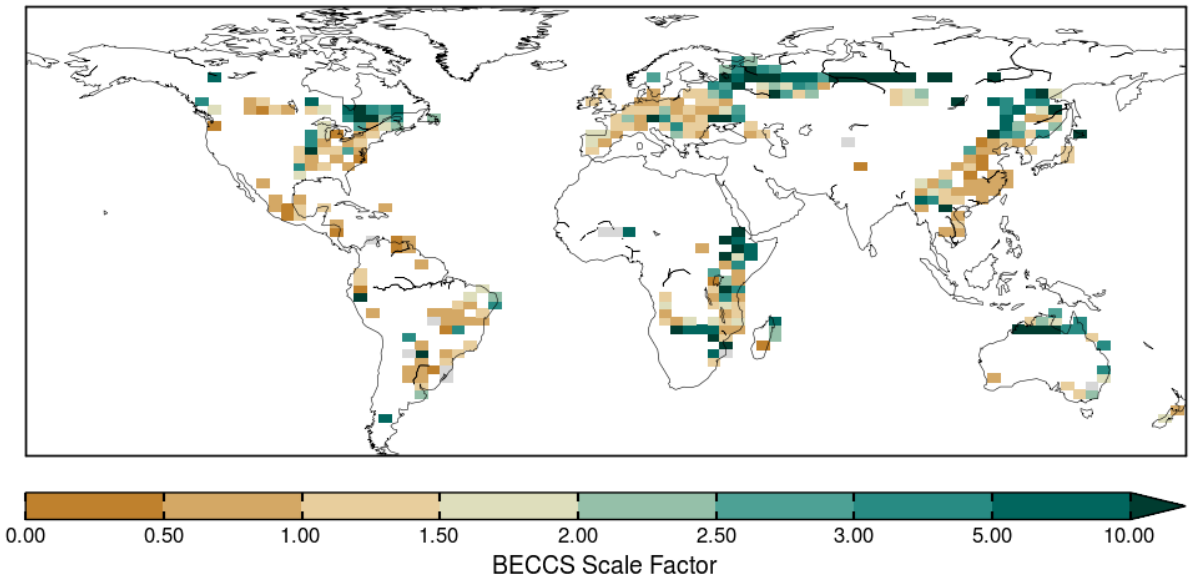


Figure 7 | Scale factor required for BECCS to be the preferable mitigation option, as opposed to natural land carbon uptake. The data represents the median of the 136 member ensemble for the optimised land-based mitigation simulation. Panel (a) is for stabilisation at 1.5°C and panel (b) is for stabilisation at 2°C.

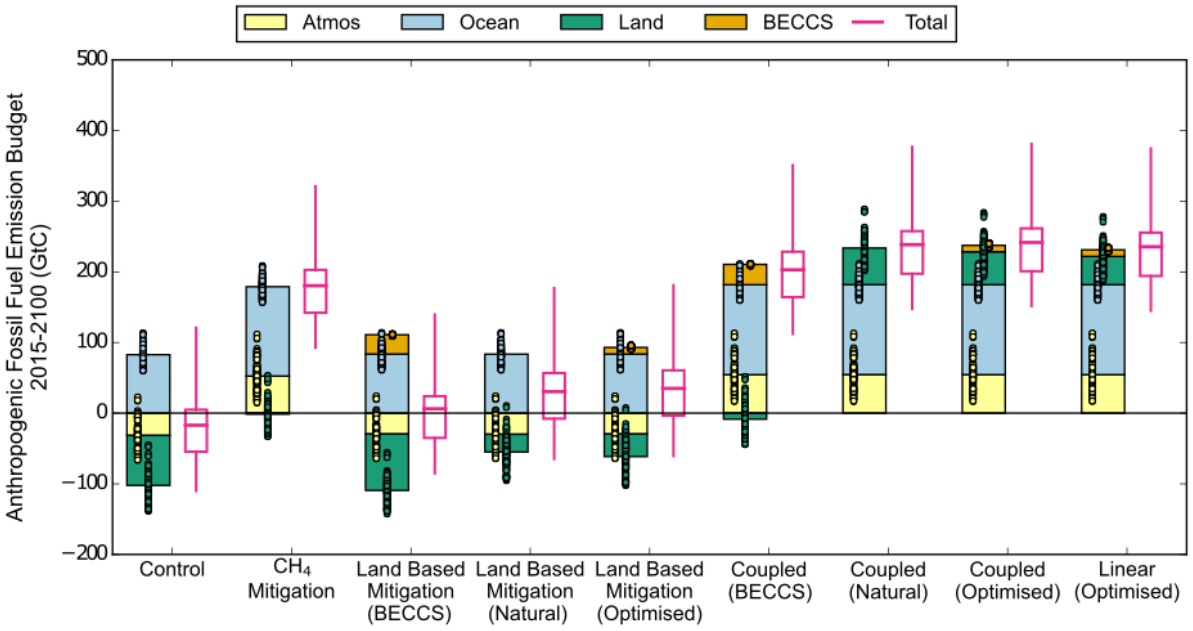
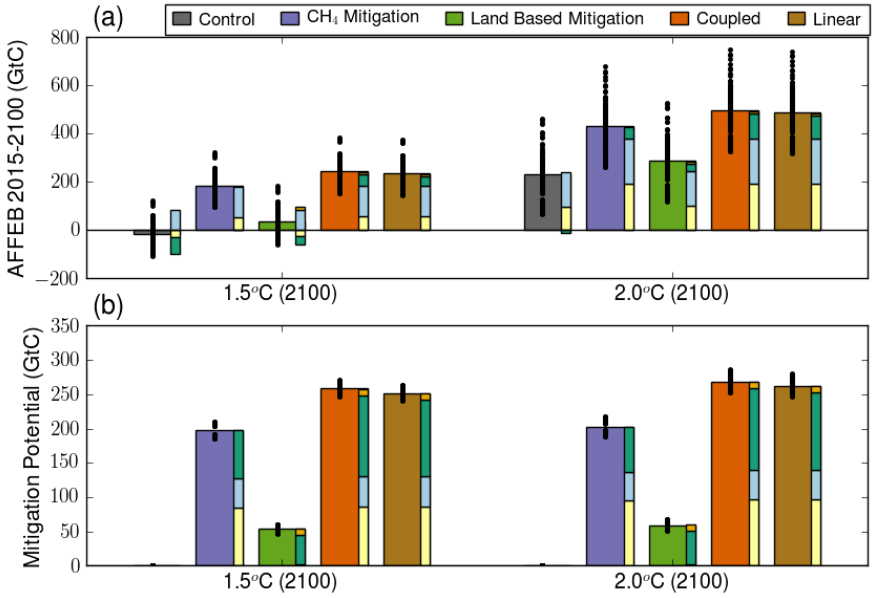


Figure 8 | The contribution to the allowable anthropogenic fossil fuel emission budget (AFFEBs, GtC) from the changes in the different carbon stores (atmosphere, ocean, land and BECCS) for the various control and mitigationscenarios, illustrated using the temperature pathways for 1.5°C of warming. The bars are the median of the component 136-member ensembles, with the individual members shown as points. The accompanying pink box and whiskers plots to the right of each set of bars are for the AFFEBs (as the sum of the changes in the component carbon stores). The box and whisker plots show the median, interquartile range, minimum and maximum derived of the resulting AFFEB ensemble. The optimised land based and coupled mitigation options selects the land use option, which maximises the AFFEB for each model grid cell. Note that the land carbon store for the CH_4 scenario is at -1.4 GtC (median of ensemble) is not visible, although the individual ensemble members can be seen as the green points.

BECCS Scale Factor (κ) = 1



BECCS Scale Factor (κ) = 3

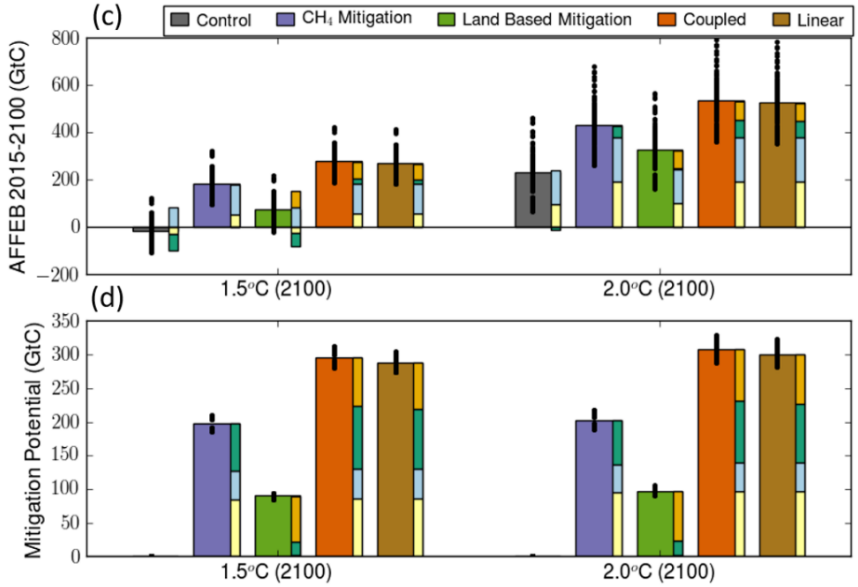


Figure 9 | Panels (a & c): The allowable anthropogenic fossil fuel emission budgets (AFFEBs; GtC) for the control (grey), CH_4 mitigation (purple), land-based mitigation (green), coupled methane and land-based mitigation (orange) and the linearly summed methane and land-based mitigation (brown), for 2 temperature pathways asymptoting at 1.5°C (left) and 2.0°C (right). (b & d) The mitigation potential (GtC) as the increase in AFFEB from the corresponding control run. The breakdown of each AFFEB and mitigation potential by the changes in the carbon stores is also shown: atmosphere (pale yellow), ocean (light blue), land (dark green) and BECCS (gold) is included alongside each bar. Note that the land carbon store for the "CH_4" scenario at -1.4 GtC (median of

ensemble) is not visible. There has however been a net increase in the land carbon store in this scenario when compared to the land carbon store in the control run (-70.8 GtC, median of ensemble).

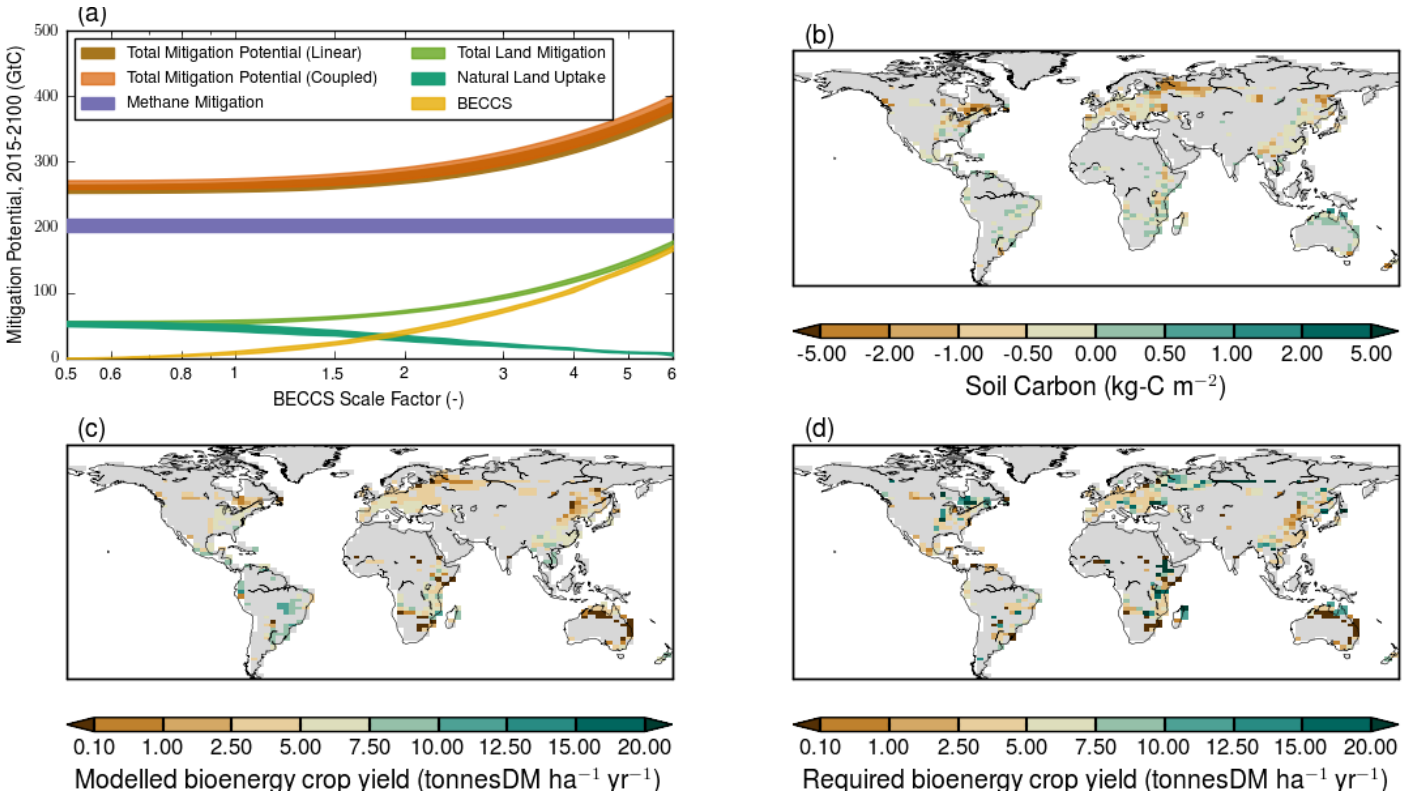


Figure 10| (a) The total and component mitigation potential (GtC) for different mitigation options, involving methane and land use, as a function of the BECCS efficiency factor (κ, Sect. 2.4.3) for the temperature pathway reaching 1.5°C. The width of the lines represent the interquartile range of the 136-member ensembles. Maps of (b) the change of the modelled soil carbon (kg-C m^{-2}) between 2015 and 2099, as the difference between the scenario with BECCS and the natural land-management scenario; (c) the modelled mean bioenergy crop yield in the JULES simulations (κ = 1) and (d) the required bioenergy crop yield for BECCS to provide a larger carbon uptake than forest regrowth/afforestation (assuming κ = κ* and 87% efficiency of BECCS). Grid cells which do not exceed 1% BECCS cover for any year in the simulation are masked grey.

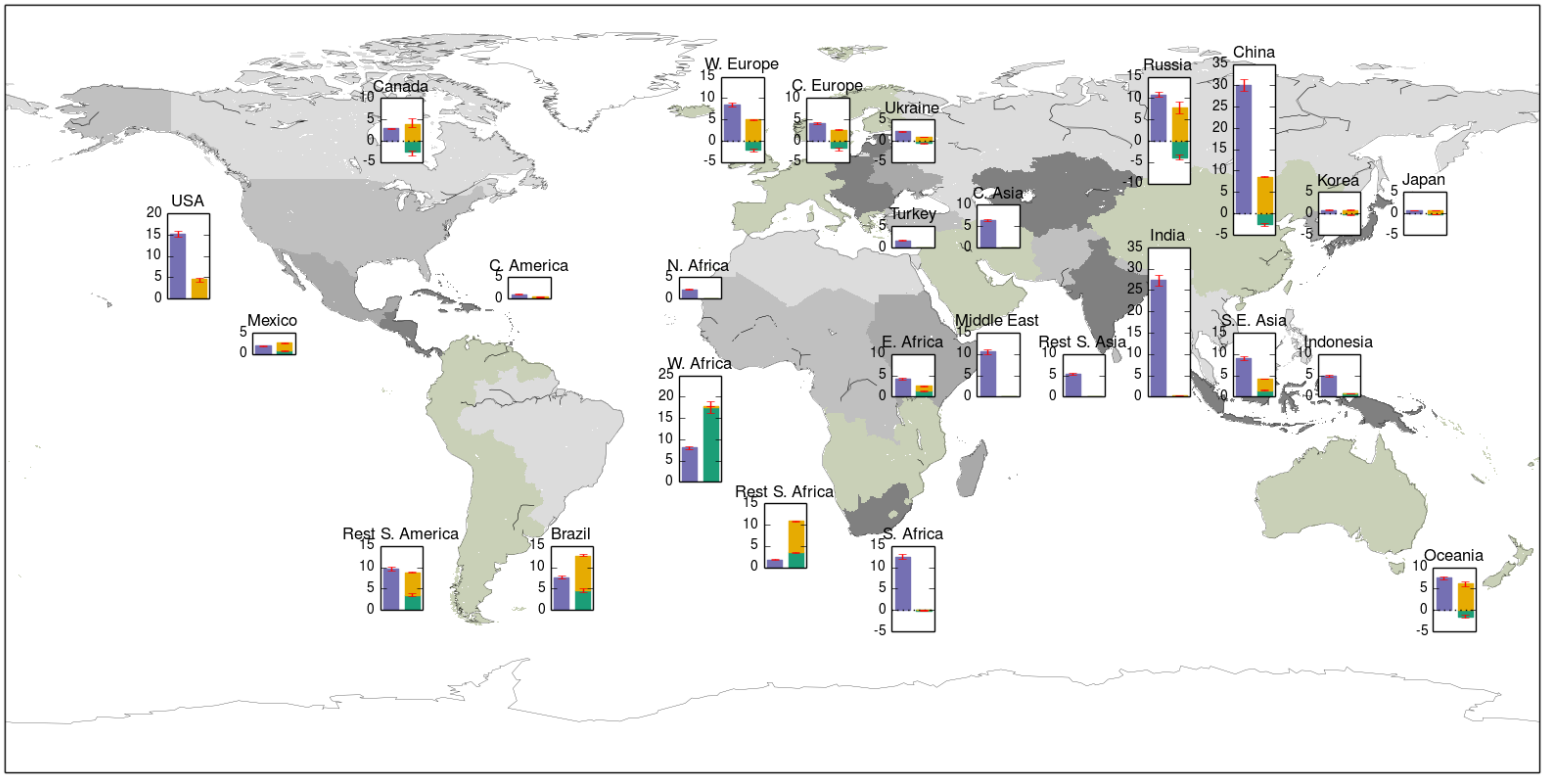
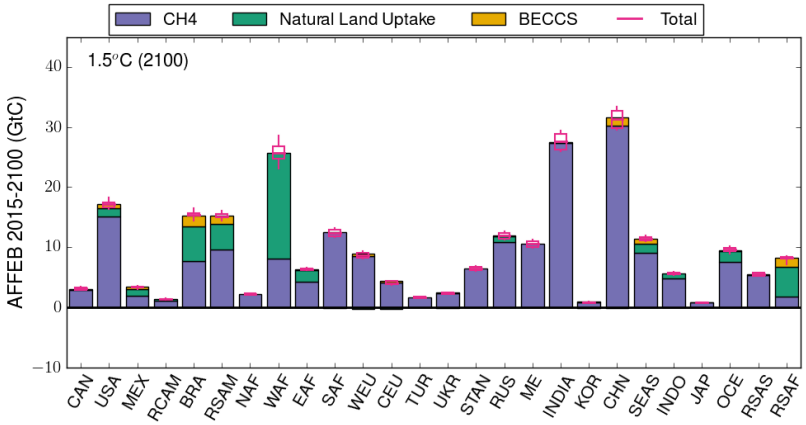
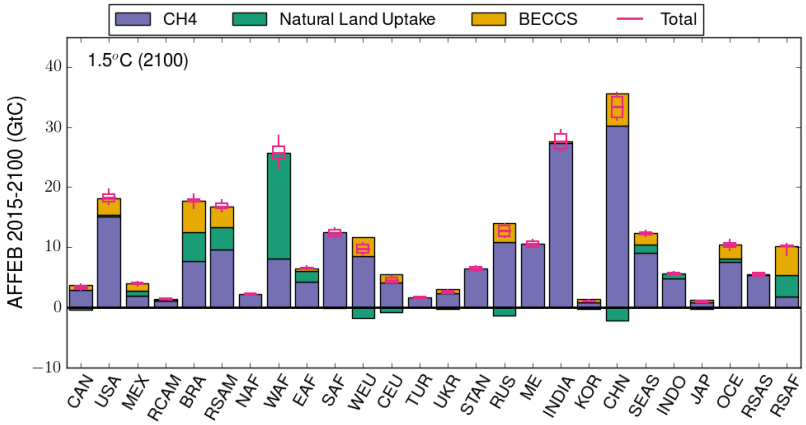


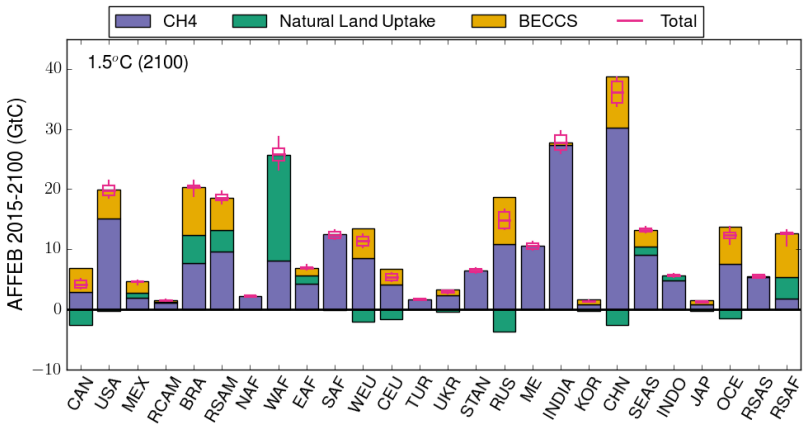
Figure 11 | The contribution to the allowable carbon emission budgets (GtC) between 2015 and 2100 for each of the 26 IMAGE IAM regions from methane mitigation (purple bars) and land-based mitigation options (green: natural land uptake; yellow: BECCS with $\kappa = 3$), for the temperature pathway stabilising at 1.5° warming without overshoot. The bars and error bars respectively show the median and the interquartile range, from the 136-member ensembles.



(a)



(b)



(c)

Figure 12 | Contribution of different mitigation options to the increase in allowable anthropogenic fossil fuel emission budgets by IMAGE region to meet the 1.5°C target. The stacked bars represent the median methane mitigation potential (purple bars) and median land-based mitigation potential (natural land uptake, green; BECCS, brown). Panel (a) is based on a BECCS scaling factor of unity, (b) a BECCS scaling factor of 2 and (c) a BECCS scaling factor of 3. The total (pink) shows the median and interquartile range of the 136-member ensembles.

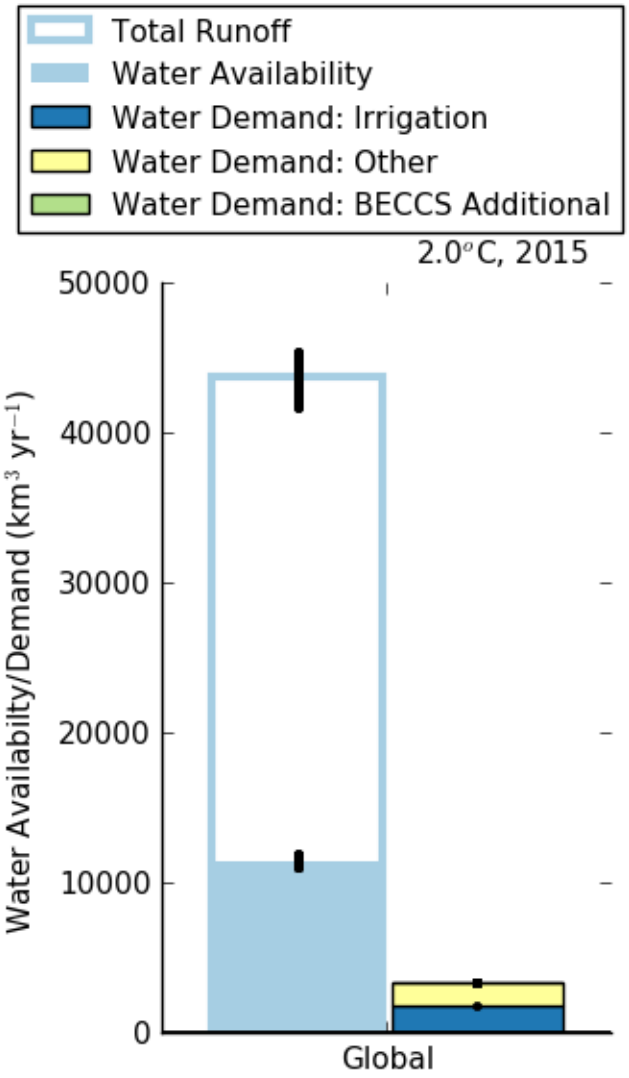


Figure 13 | Global water availability (filled light blue bar) as a regionally dependent fraction of runoff (hollow light blue bar) for the year 2015. The water demand for irrigation (dark blue) and for other uses (i.e., energy generation, industry and domestic; yellow), are taken from the SSP2-RCP2.6-IMAGE database. Note there is very little BECCS additional water demand (green) in 2015.

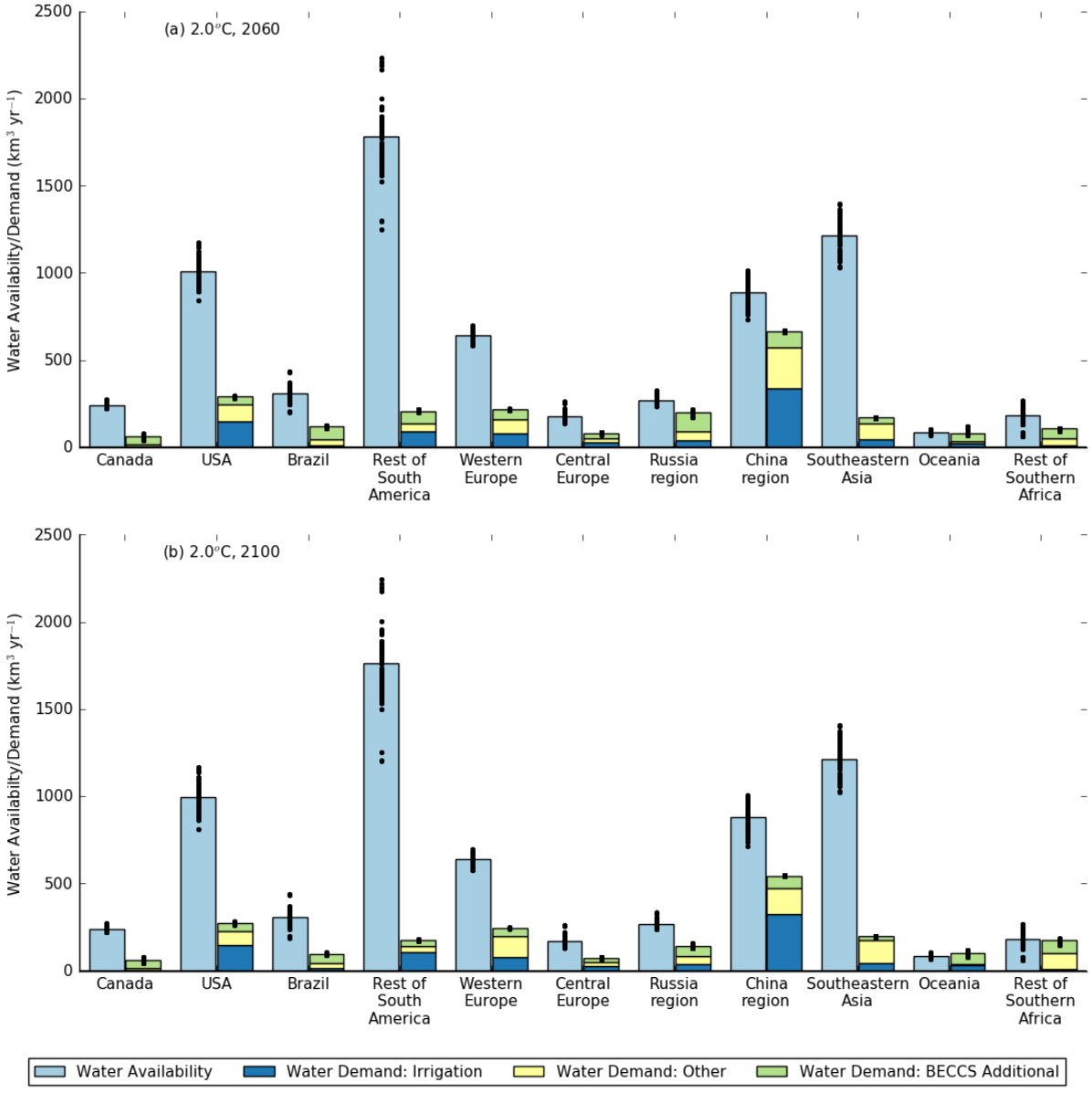


Figure 14 | Water availability (light blue), SSP2-IMAGE water demand estimates for irrigation (dark blue), other uses (i.e., energy generation, industry and domestic; yellow) and the additional water demand from BECCS (green) for the years 2059-2060 and 2099-2100 for the 2.0°C warming target, with a BECCS κ factor of 3. The points are the individual results from the 136-member ensembles, while the bars are the corresponding median values of the ensembles.

Tables

Table 1 | The IMOGEN-JULES and post processing scenario runs, key features and the input and prescribed datasets used in the scenarios.

(a) IMOGEN-JULES modelling scenarios (Note 1)

	Scenario (Abbreviation) **Key features of the Scenario**	**Scenario-specific input and prescribed datasets (Notes 2, 3)**
1.	**Control ("CTL")** • Agricultural land accrued to feed growing populations associated with the SSP2 pathway • No deployment of BECCS • Anthropogenic CH_4 emissions rise from 318 Tg yr^{-1} in 2005 to 484 Tg yr^{-1} in 2100 • Effects of the methane and carbon-climate feedbacks from wetlands and permafrost thaw included	Scenario-specific input data • Time series of radiative forcing by non-CO_2 GHG and other non-CO_2 climate forcers, for the IMAGE SSP2 baseline scenario • Time series of annual global atmospheric concentrations of CH_4 and N_2O for the IMAGE SSP2 baseline scenario Scenario-specific prescribed data • Gridded annual time series of areas assigned to agriculture (crops & pasture), for the IMAGE SSP2 baseline scenario, converted into fractions of the IMOGEN-JULES grid cell
2.	**Methane mitigation ("CH_4")** • Agricultural land-use as in Control ("CTL") scenario • Anthropogenic CH_4 emissions decline from 318 Tg yr^{-1} in 2005 to 162 Tg yr^{-1} in 2100, from the IMAGE SSP2 RCP1.9 scenario • Effects of the methane and carbon-climate feedbacks from wetlands and permafrost thaw included	Scenario-specific input data • Time series of radiative forcing by non-CO_2 GHG and other non-CO_2 climate forcers, for the IMAGE SSP2 RCP1.9 scenario • Time series of annual global atmospheric concentrations of CH_4 and N_2O for the IMAGE SSP2 RCP1.9 scenario Scenario-specific prescribed data • As 1, gridded annual time series of area assigned to agriculture (crops & pasture). Converted into fractions of the IMOGEN-JULES grid cell
3.	**Land-based mitigation, including BECCS ("BECCS")** • Land use change based on the IMAGE SSP2 RCP1.9 scenario • High levels of REDD and full reforestation • Food-first policy so that bioenergy crops (BE) are only implemented on land not required for food production • Anthropogenic CH_4 emissions as in Control ("CTL") scenario • Effects of the methane and carbon-climate feedbacks from wetlands and permafrost thaw included	Scenario-specific input data • Time series of radiative forcing by non-CO_2 GHG and other non-CO_2 climate forcers, for the IMAGE SSP2 baseline scenario • Time series of annual global atmospheric concentrations of CH_4 and N_2O for the IMAGE SSP2 baseline scenario (as used in "CTL") Scenario-specific prescribed data • Gridded annual time series of areas assigned to agriculture (crops & pasture) and within that the area for bioenergy crops, for the IMAGE SSP2 RCP1.9 scenario. Converted into a fraction of the IMOGEN-JULES grid cell
4.	**Land-based mitigation with no BECCS ("Natural")** • Land use as 3, except any land area allocated to bioenergy crops is set to zero, allowing expansion of natural vegetation • Anthropogenic CH_4 emissions as in Control ("CTL") scenario • Effects of the methane and carbon-climate feedbacks from wetlands and permafrost thaw included	Scenario-specific input data • Time series of radiative forcing by non-CO_2 GHG and other non-CO_2 climate forcers, for the IMAGE SSP2 baseline scenario • Time series of annual global atmospheric concentrations of CH_4 and N_2O for the IMAGE SSP2 baseline scenario (as used in "CTL") Scenario-specific prescribed data • Gridded annual time series of areas assigned to agriculture (crops & pasture). As 3, except any land allocated to bioenergy crops is set to zero. Converted into a fraction of the IMOGEN-JULES grid cell

5.	**Combined methane & land-based mitigation** **"Coupled(CH_4+BECCS)"** • Combines CH_4 mitigation of 2 with land-based mitigation scenario of 3	Scenario-specific input data • As 2, time series of radiative forcing by non-CO_2 GHG and other non-CO_2 climate forcers, for the IMAGE SSP2 RCP1.9 scenario • As 2, time series of annual global atmospheric concentrations of CH_4 and N_2O for the IMAGE SSP2 RCP1.9 scenario Scenario-specific prescribed data • As 3, gridded annual time series of areas assigned to agriculture (crops & pasture) and within that the area for bioenergy crops, for the IMAGE SSP2 RCP1.9 scenario. Converted into prescribed fractions of the IMOGEN-JULES grid cell
6. C	**Combined methane & land-based mitigation with no BECCS** **"Coupled (CH_4+Natural)"** • Combines CH_4 mitigation of 2 with land-based mitigation scenario of 4	Scenario-specific input data • As 2, time series of radiative forcing by non-CO_2 GHG and other non-CO_2 climate forcers, for the IMAGE SSP2 RCP1.9 scenario • As 2, time series of annual global atmospheric concentrations of CH_4 and N_2O for the IMAGE SSP2 RCP1.9 scenario Scenario-specific prescribed data • As 4, gridded annual time series of areas assigned to agriculture (crops & pasture). Converted into a fraction of the IMOGEN-JULES grid cell

(b) Post-processing scenarios (Note 1)

	Scenario **"Abbreviation"**	**Description of the Scenario**
7.	**Optimisation of land-based mitigation** **"Land-based mitigation: Optimised"**	• Optimisation of scenarios 3 and 4 by selecting the scenario which has the larger carbon uptake, on a grid cell by grid cell basis
8.	**Optimisation of the combined methane & land-based mitigation** **"Coupled Optimised"**	• Optimisation of scenarios 5 and 6 by selecting the scenario which has the larger carbon uptake, on a grid cell by grid cell basis

Notes

1. Each scenario comprises two 136-member ensembles (34 GCMs x 2 ozone damage sensitivities x 2 methanogenesis Q_{10} temperature sensitivities), one for the 1.5°C warming target and the second for the 2°C warming target.
2. All of the above scenarios also use time series of (1) observed temperature changes between 1850 and 2015; (2) profiles of temperature change between 2015 and 2100 to achieve the 1.5°C and the 2°C warming targets; and (3) the radiative forcing changes of non-CO_2 radiative forcing between 1850 and 2015.
3. We define (a) a "prescribed" dataset as one that is used unchanged in the IMOGEN-JULES modelling; (b) an "input" dataset as one that provides the initial values that are subsequently changed.

Table 2 | IMAGE regions, the maximum area of BECCS deployed (Mha) and the main differences in land use between the BECCS and Natural scenarios.

Region	Abbreviation	Max. area of bioenergy crops (Mha)	Main land-use difference between BECCS and Natural scenarios
Canada	CAN	65.9	Forest to BECCS in BECCS scenario
USA	USA	39.0	Agricultural land and forest to BECCS (BECCS). Agricultural land to forest (Natural)
Mexico	MEX	7.1	Agricultural land to BECCS and forest (BECCS). Agricultural land to forest (Natural)
Central America	RCAM	0.5	Little BECCS. Agricultural land to forests in both scenarios.
Brazil	BRA	27.8	Agricultural land to BECCS and forest (BECCS). Agricultural land to forest (Natural)
Rest of South America	RSAM	20.3	Agricultural land to BECCS and forest (BECCS). Agricultural land to forest (Natural)
Northern Africa	NAF	0.0	No BECCS. No real differences between scenarios
Western Africa	WAF	3.1	Little BECCS. Agricultural land to forests in both scenarios.
Eastern Africa	EAF	33.9	Agricultural land to BECCS and forest (BECCS). Agricultural land to forest (Natural)
South Africa	SAF	1.0	Little BECCS. Agricultural land to forests in both scenarios.
Rest of Southern Africa	RSAF	63.7	Agricultural land to BECCS and forest (BECCS). Agricultural land to forest (Natural)
Western Europe	WEU	23.6	Forest to BECCS in BECCS scenario
Central Europe	CEU	19.3	Forest to BECCS in BECCS scenario
Turkey	TUR	0.0	No BECCS. No real differences between scenarios
Ukraine Region	UKR	11.4	Forest to BECCS in BECCS scenario
Central Asia	STAN	0.7	Little BECCS. No real differences between scenarios
Russia Region	RUS	146.1	Forest to BECCS in BECCS scenario
Middle East	ME	0.0	No BECCS. No real differences between scenarios
India	INDIA	6.0	Forest to BECCS in BECCS scenario
Korea Region	KOR	4.3	Forest to BECCS in BECCS scenario
China	CHN	58.1	Forest to BECCS in BECCS scenario
South East Asia	SEAS	24.5	Forest to BECCS in BECCS scenario. Agricultural land to forest (Natural)
Indonesia	INDO	0.0	No BECCS. Agricultural land to forests in both scenarios.
Japan	JAP	2.7	Forest to BECCS in BECCS scenario
Rest of South Asia	RSAS	0.0	No BECCS. No real differences between scenarios
Oceania	OCE	78.7	Forest to BECCS in BECCS scenario

Table 3 | For the 1.5°C temperature profile, the mean of the 34-GCM member ensembles for the "CTL" and mitigation scenarios for the different factorial runs (low Q_{10}/low O_3, low Q_{10}/high O_3, high Q_{10}/low O_3 and high Q_{10}/high O_3), the standard deviation of the full 136-member ensemble (GtC), the derived standard deviations for land processes (σ_{land}) and climate ($\sigma_{climate}$, as represented by the 34 GCMs) and the ratio of $\sigma_{climate}/\sigma_{land}$ for (a) the Anthropogenic Fossil Field CO_2 Emission Budgets and (b) the Mitigation Potential (= scenario – CTL).

(1) AFFEB

Scenario	Mean of 34-member Factorial Run (GtC)				Standard Deviation (GtC)			Ratio
	Low Q_{10} Low O_3	Low Q_{10} High O_3	High Q_{10} Low O_3	High Q_{10} High O_3	136-member Ensemble	Land σ_{land}	Climate $\sigma_{climate}$	$\sigma_{climate}$:σ_{land}
CTL	-9.66	-20.58	-18.91	-31.06	47.12	7.60	46.50	6.12
CH_4	179.44	186.79	168.73	174.90	47.54	6.59	47.08	7.14
BECCS	6.49	3.42	-2.09	-5.80	47.45	4.76	47.21	9.91
Natural	42.57	24.60	35.00	16.05	48.95	10.07	47.90	4.75
Optimised Land-based	46.42	29.18	37.89	20.00	48.85	9.84	47.85	4.86
Linear BECCS+CH_4	195.58	210.79	185.55	200.15	48.64	9.07	47.79	5.27
Linear_Natural+CH_4	231.67	231.97	222.64	222.00	48.70	4.76	48.47	10.19
Linear optimised	235.51	236.55	225.53	225.96	48.69	5.16	48.42	9.39
Coupled BECCS+CH_4	199.69	214.62	189.50	203.94	48.48	9.01	47.64	5.29
Coupled Natural+CH_4	237.83	238.95	228.72	228.91	48.60	4.80	48.36	10.07
Coupled optimised	241.50	243.29	231.35	232.60	48.60	5.27	48.31	9.17

(2) Mitigation Potential

Scenario	Mean of 34-member Factorial Run (GtC)				Standard Deviation (GtC)			Ratio
	Low Q_{10} Low O_3	Low Q_{10} High O_3	High Q_{10} Low O_3	High Q_{10} High O_3	136-member Ensemble	Land σ_{land}	Climate $\sigma_{climate}$	$\sigma_{climate}$:σ_{land}
CTL	-	-	-	-	-	-	-	-
CH_4	189.10	207.37	187.64	205.96	9.28	9.18	1.39	0.15
BECCS	16.14	24.01	16.82	25.26	4.24	4.11	1.05	0.26
Natural	52.23	45.18	53.91	47.11	3.93	3.58	1.62	0.45
Optimised Land-based	56.07	49.76	56.80	51.06	3.44	3.06	1.57	0.51
Linear BECCS+CH_4	205.24	231.38	204.46	231.21	13.39	13.23	2.09	0.16
Linear_Natural+CH_4	241.33	252.55	241.55	253.06	6.14	5.69	2.32	0.41
Linear optimised	245.17	257.13	244.44	257.02	6.55	6.14	2.28	0.37
Coupled BECCS+CH_4	209.34	235.20	208.41	235.00	13.27	13.12	2.01	0.15
Coupled Natural+CH_4	247.48	259.54	247.63	259.97	6.49	6.10	2.21	0.36
Coupled optimised	251.15	263.87	250.26	263.66	6.89	6.54	2.17	0.33

Table 4a | Comparison by IMAGE region of the modelled available water ($km^3\ yr^{-1}$), the projected water withdrawals ($km^3\ yr^{-1}$) for irrigation and for other anthropogenic activities (energy generation, industry, domestic) from the IMAGE SSP2-RCP2.6 scenario, and the additional water required for BECCS ($km^3\ yr^{-1}$ and as percentages of the net available water and of the water withdrawals for irrigation and other), for the year 2060. The percentage of runoff available for human use by IMAGE region is also included.

Region	Abbreviation	% of Regional Runoff Available	Available Water ($km^3\ yr^{-1}$)	Water Demand			Total Demand as % of Available Water	BECCS Demand as % of Total Demand
				Irrigation ($km^3\ yr^{-1}$)	Other ($km^3\ yr^{-1}$)	BECCS ($km^3\ yr^{-1}$)		
Canada	CAN	40%	243.19	3.39	14.21	44.45	25.5%	71.6%
USA	USA	5%	1,010.82	149.55	96.07	44.55	28.7%	15.4%
Mexico	MEX	5%	75.89	76.58	25.56	24.48	166.8%	19.3%
Central America	RCAM	5%	185.92	8.16	15.49	2.28	13.9%	8.8%
Brazil	BRA	40%	310.65	12.24	34.44	73.12	38.6%	61.0%
Rest of South America	RSAM	5%	1,779.42	93.50	46.49	67.66	11.7%	32.6%
Northern Africa	NAF	5%	0.11	61.60	54.63	0.00	-	-
Western Africa	WAF	5%	1,962.47	28.29	118.83	0.39	7.5%	0.3%
Eastern Africa	EAF	5%	485.18	53.92	63.10	2.45	24.6%	2.1%
South Africa	SAF	5%	0.60	13.45	9.28	0.48	3868.3%	2.1%
Rest of Southern Africa	RSAF	5%	182.48	10.03	41.36	56.02	58.9%	52.2%
Western Europe	WEU	5%	642.34	78.72	82.01	56.22	33.8%	25.9%
Central Europe	CEU	5%	176.27	27.46	22.32	29.68	45.1%	37.4%
Turkey	TUR	5%	29.98	60.35	15.86	0.00	-	-
Ukraine Region	UKR	5%	67.47	11.73	25.90	12.28	74.0%	24.6%
Central Asia	STAN	5%	20.57	88.26	32.62	0.00	-	-
Russia Region	RUS	40%	270.32	42.30	51.60	103.87	73.2%	52.5%
Middle East	ME	5%	8.65	149.55	40.97	0.00	-	-
India	INDIA	5%	319.36	374.18	501.06	0.00	-	-
Korea Region	KOR	5%	42.85	6.20	9.75	12.64	66.7%	44.2%
China	CHN	5%	887.26	338.81	236.89	87.73	74.8%	13.2%
South East Asia	SEAS	5%	1,212.00	46.52	92.99	31.56	14.1%	18.4%
Indonesia	INDO	5%	1,293.05	8.18	113.87	0.00	-	-
Japan	JAP	5%	209.49	2.79	18.99	7.69	14.1%	26.1%
Rest of South Asia	RSAS	5%	74.57	259.95	154.42	0.00	-	-
Oceania	OCE	5%	85.46	24.99	8.91	48.06	95.9%	58.6%

Table 4b | As Table 4a for 2100

Region	Abbreviation	% of Regional Runoff Available	Available Water (km^3 yr^{-1})	Water Demand			Total Demand as % of Available Water	BECCS Demand as % of Total Demand
				Irrigation (km^3 yr^{-1})	Other (km^3 yr^{-1})	BECCS (km^3 yr^{-1})		
Canada	CAN	40%	240.14	4.31	11.72	45.21	25.5%	73.8%
USA	USA	5%	993.09	148.57	81.35	45.45	27.7%	16.5%
Mexico	MEX	5%	72.79	77.27	23.78	11.14	154.1%	9.9%
Central America	RCAM	5%	182.12	8.74	13.96	0.66	12.8%	2.8%
Brazil	BRA	40%	307.53	12.31	30.80	54.89	31.9%	56.0%
Rest of South America	RSAM	5%	1,765.14	103.97	38.34	32.65	9.9%	18.7%
Northern Africa	NAF	5%	0.11	57.89	56.98	0.00	-	-
Western Africa	WAF	5%	1,953.10	37.23	262.07	0.62	15.4%	0.2%
Eastern Africa	EAF	5%	485.02	58.96	128.33	20.54	42.8%	9.9%
South Africa	SAF	5%	0.60	13.43	7.50	0.45	3563.3%	2.1%
Rest of Southern Africa	RSAF	5%	179.63	11.20	89.87	74.85	97.9%	42.5%
Western Europe	WEU	5%	637.68	80.39	118.64	45.25	38.3%	18.5%
Central Europe	CEU	5%	171.05	26.90	20.63	23.19	41.3%	32.8%
Turkey	TUR	5%	29.52	60.49	12.87	0.00	-	-
Ukraine Region	UKR	5%	66.45	10.40	19.58	8.62	58.1%	22.3%
Central Asia	STAN	5%	19.67	82.08	37.90	0.00	-	-
Russia Region	RUS	40%	266.36	40.25	43.82	58.40	53.5%	41.0%
Middle East	ME	5%	8.60	136.63	39.30	0.00	-	-
India	INDIA	5%	320.08	388.69	585.48	0.00	-	-
Korea Region	KOR	5%	42.73	7.41	5.47	0.00	-	-
China	CHN	5%	881.00	326.62	144.80	72.75	61.8%	13.4%
South East Asia	SEAS	5%	1,213.01	45.46	131.95	19.49	16.2%	9.9%
Indonesia	INDO	5%	1,291.53	15.08	114.33	0.00	-	-
Japan	JAP	5%	208.43	2.12	13.29	6.94	10.7%	31.1%
Rest of South Asia	RSAS	5%	74.19	245.78	227.85	0.00	0.0%	0.0%
Oceania	OCE	5%	85.46	30.57	8.77	62.96	136.5%	160.0%