Producing Policy-relevant Science by Enhancing Robustness and Model Integration for the Assessment of Global Environmental Change

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ABSTRACT

We use the flexible model coupling technology known as the bespoke framework generator to link established existing modules representing dynamics in the global economy (GEMINI-E3), the energy system (TIAM-WORLD), the global and regional climate system (MAGICC6, PLASIM-ENTS and ClimGEN), the agricultural system, the hydrological system and ecosystems (LPJmL), together in a single integrated assessment modelling (IAM) framework, building on the pre-existing framework of the Community Integrated Assessment System. Next, we demonstrate the application of the framework to produce policy-relevant scientific information. We use it to show that when using carbon price mechanisms to induce a transition from a high-carbon to a low-carbon economy, prices can be minimised if policy action is taken early, if burden sharing regimes are used, and if agriculture is intensified. Some of the coupled models have been made available for use at a secure and user-friendly web portal.

1. Introduction

Integrated assessment models are increasingly used as tools for projecting scenarios of global change by drawing together information from a variety of disciplines. However, such models often do not assemble detailed treatments of both the earth system and the global economy within a single framework, and often consist of single pieces of software. Here we describe the assembly and use of a modular integrated assessment framework that is based on the principle of coupling together alternative combinations of modules, each implemented at a different institution, to produce an enhanced integrated modelling framework (Warren et al., 2013; http://ermitage.cs.man.ac.uk). We couple together state-of-the-art, intermediately complex models representing the global economy and social actors within it, the physical climate system, the energy system, the agricultural system, the hydrological system, and ecosystems. This type of integrated assessment modelling is needed in order to study the complex interactions between climate change, climate change impacts, climate change mitigation, and decisions about land use management. The work was performed as part of the EU project 'Enhancing robustness and model integration for...
the assessment of global environmental change (ERMITE). These integrated assessments are now of particular topical interest in view of the recent adoption of the United Nations Framework on Climate Change’s Paris Agreement (UNFCCC, 2015) by 195 countries.

Most of this framework is incorporated within the Community Integrated Assessment System (CIAS) (Warren et al., 2008), whilst some of the coupled models exist independently of CIAS (specifically, the coupling between the energy technology model REMIND and the land use allocation model MagPie, see Table 2 for details). The approach is based on the advanced flexible bespoke framework generator which is language-independent (Armstrong et al., 2009). The flexible approach allows new modules to be added to the system with minimum disruption, for example when climate models are upgraded with new information, or when updated modules become available simulating climate impacts in new sectors. The approach has created long-lasting coupled models available for use in research for the future, by drawing together a range of models created at different institutions. The co-location of many of the models in the same system (CIAS) allows for increased, easy use of the models in the future at a secure and user-friendly web portal.

2. Methodology

The first step in the modelling processes is to determine conceptually the required linkages between models. This was achieved through bilateral discussions at workshops which allowed model developers from different disciplines to work together. Initially the team created a prioritised list of model couplings needed to answer the research questions we have. Once the list of model couplings had been agreed, the team then worked together to determine the scientific requirements of the couplings. These requirements included detailing (a) which are the variables output from one model that are to become the input to another model? (b) are any unit conversions required? (c) is any spatial or temporal aggregation required to allow for differences in the spatial or temporal resolution used in different models? (d) when during the operation of the code should the variables be passed? Once these requirements had been determined we used the Bespoke Framework Generator version 2 (BFG2) to couple models together according to the requirements in a language-independent fashion (Ford et al., 2006; Warren et al., 2008). BFG2 has a simple interface which allows users to automatically create metadata describing model linkages; and it continues by using this meta-data to automatically generate the coupling code. The metadata follows a ‘DCD’ approach: it contains Description (D) information about the variables to be exchanged between the models that are to be coupled, specifically the variable names, units, and temporal and spatial scales; Composition (C) information detailing which quantities should be exchanged between the model codes at which times during the running of the code; and Deployment (D) information detailing which machines will run the code. We initially coupled pairs of models together before moving on to more complex coupled models involving three or more components. Finally, couplings were incorporated into framework of the Community Integrated Assessment System (CIAS, Fig. 1), which allows users to execute the couplings at a user-friendly web portal.

CIAS (Warren et al., 2008) is a framework that supports and enables the creation and running of integrated assessment models. It connects together alternative sets of component models: thus one of these sets is broadly equivalent to ‘an integrated assessment model’ and may be referred to as ‘a coupled model’. It is flexible and multi-modal, and enables models to communicate with each other even if they are written in different programming languages or operate on different platforms. The CIAS web portal supports users in running the integrated models: it is facilitated by the softIAM technology (Goswami and Warren, 2012). For each coupling, the softIAM technology supports a variety of coupling-specific features related to the selection of modes of operation, changing model parameters, selecting variables for output, and user management. Model coupling outputs are stored in a database, and can be accessed from the web portal. Table 1 provides the list of models used, and Table 2 shows the list of linkages between the modules which we created.

We used a third key software technology, statistical emulation, to speed up the run time of some of our model couplings. In this approach, a model is replaced by a computationally much faster and functionally smoother model ‘emulator’, derived from a large ensemble of simulations. We created emulators for PLASIM-ENTS (Holden et al., 2014) and also for the simulation of net primary production and crop yields by LPJmL (Oyebamiji et al., 2015). The methodologies are described in detail in these references. In summary, the PLASIM-ENTS emulator uses singular vector decompositions of the spatiotemporal outputs of a large ensemble of transient 21st century climate simulations, considering a wide range of future emissions scenarios. The dominant components of the decompositions are fitted as polynomial functions of future forcing and model parameters. The approach represents an advance on pattern scaling as it allows us to address non-linear spatiotemporal feedbacks and model parametric uncertainty by representing multiple modes of variability. The LPJmL emulator is constructed in a two-stage approach. The first stage uses step-wise regression to fit crop yields as smooth functions of local climate variables, under the assumption that each LPJmL grid cell is an independent sample. The second stage combines principal component analysis and weighted least squares to allow for bias in predicted spatial patterns, correcting for the anticipated residual of the first stage. In Table 2, coupling sequences in which the models have suffix ‘em’ refers to emulators of the full codes.

Fig. 2 illustrates the emulation of precipitation from PLASIM-ENTS. Deriving the emulated precipitation fields required ∼1 min of CPU time, compared to ∼1 year of computer time required for the full simulation. This can result in the loss of representation of more complex processes such as feedbacks and non-linearities, which might be important. Thus, use of emulators of more complex models allows the statistical (as opposed to mechanistic) representation of more complex processes than would otherwise be possible within integrated models. The statistical emulation needs to be robust: in this example the emulated ensemble reproduces the simulated mean field extremely closely in relation to the ensemble variance. It is also able to reproduce the pattern of the simulated uncertainty field, though somewhat understating its magnitude.

3. Example couplings

3.1. Example 1: PLASIM-ENTsem_ClimGEN_LPJmLem

In this relatively simple coupling (Fig. 3a), measures of global climate change such as temperature are used to drive a pattern scaling module ClimGEN which in turn drives an emulator of a climate change impact model, LPJmL. The process begins with the provision of historical and projected global time series of greenhouse gas concentrations to the climate model emulator PLASIM-ENTS (Holden et al., 2014) which simulates global climate changes for near-surface temperature, precipitation and cloud cover on a 5° grid scale. The seasonally-resolved climate projections are passed to ClimGen which down-scales the data to a 0.5° grid. In pattern scaling, linear relationships between projected local climate change and projected global mean temperature change are diagnosed directly from outputs of global circulation models; these are combined with observed climatological data to create projected fields of climate change (here precipitation and temperature) at a resolution of 0.5° x 0.5° (Warren et al., 2008 for further detail). Finally, the downscaled climate change projections are used by LPJmLem to project impacts resulting from the studied global climate change scenarios. Outputs from this coupling are, for example, gridded projections of crop yields.
3.2. Example 2: GEMINI-E3_PLASIM-ENTSem

This particular coupling (Fig. 3b) has been designed to use the emulator of the climate model PLASIM-ENTSem to create greenhouse gas emissions constraints for the macro-economic model GEMINI-E3 in order to derive climate policy (such as a carbon tax scheme) that constrains the global annual mean temperature rise occurring between pre-industrial times and 2050 to a particular level. It is also designed to investigate the impacts of climate changes on heating and cooling demands, and the economic consequences thereof.

Since GEMINI-E3 is a time-step optimization model, it is not feasible to compute endogenously an optimal emissions path with respect to the economy. For this reason, we have implemented a soft coupling approach, in which no optimisation occurs, which gives realistic emissions profiles given the anticipated temperature expectations. These emissions profiles are used in GEMINI-E3 as an upper bound on the emissions for the assessment of potential climate policies. As the number of “satisfactory” emissions trajectories is potentially unlimited, the coupling constrains its search to a subset of trajectories with two functional forms - a class of simple linear functions, and a class of more complex smooth polynomials. For each proposed trajectory, PLASIM-ENTSem can compute a temperature increase and the coupling algorithm selects the one that meets the given warming target. For the resulting selected trajectory, PLASIM-ENTSem also provides Heating Degree Days (HDD) and Cooling Degree Days (CDD) to GEMINI-E3. This allows GEMINI-E3 to evaluate the impact of climate change on heating and cooling demands and the resultant economic consequences.

Outputs of this coupling are economic measures for each economic region in each time period (e.g. discounted and total welfare); permit allocation; GDP; carbon taxes; and the heating and cooling demand.

Table 1

<table>
<thead>
<tr>
<th>Type of model</th>
<th>Model</th>
<th>Brief description and key reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Overall Integrator</td>
<td>CIAS (UEA)</td>
<td>Community Integrated Assessment System: links combinations of models together in a flexible fashion to address policy questions (Warren et al., 2008)</td>
</tr>
<tr>
<td>Global welfare, energy and technology</td>
<td>REMIND-R</td>
<td>An inter-temporal optimization model maximizing global welfare subject to equilibrium conditions on different markets (Leimbach et al., 2010; Luderer et al., 2015).</td>
</tr>
<tr>
<td>Global macro-economic</td>
<td>GEMINI-E3</td>
<td>A large-scale, global CGE model, covers around 20 regions at World level (with explicitly EU, USA, India, China). It has a disaggregation of industries and types of inputs that is specifically designed to allow for substitution in energy production and use. GEMINI-E3 has been used extensively to simulate national and international climate policies (<a href="http://gemini-e3.epfl.ch/">http://gemini-e3.epfl.ch/</a>). (Bernard and Vielle, 2003, 2008)</td>
</tr>
<tr>
<td>Land use allocation</td>
<td>MAgPIE</td>
<td>Demand in 10 categories of food and feed energy is simulated in 10 economic world regions, and is met by 20 cropping activities and 3 livestock activities. Trade in food products between regions is simulated endogenously. Coupled to the grid-based dynamic model LPjML to simulate spatially explicit land-use and water-use patterns whilst considering technological and agro-economic change, including trade. (Lotze-Campen et al., 2008; Popp et al., 2014)</td>
</tr>
<tr>
<td>Energy and technology (World)</td>
<td>TIAM-WORLD</td>
<td>A technology-rich model of the entire energy/emission system of the World split into 16 regions, providing a detailed representation of the procurement, transformation, trade, and consumption of a large number of energy forms. (Loulou and Labriet, 2008; Loulou et al., 2009; Labriet et al., 2012, 2013a, 2013b).</td>
</tr>
<tr>
<td>Global Climate</td>
<td>Magicc-6 (UEA)</td>
<td>Simple, widely used climate model tuned to emulate alternative complex global circulation models. Can simulate global climate change outcomes for the RCP scenarios. See wiki.magicc.org/index.php?title=The_MAGICC_Wiki. (Meinshausen et al., 2011).</td>
</tr>
<tr>
<td>Global climate</td>
<td>PLASIM-ENTSem</td>
<td>An emulator of an intermediate complexity Global Climate Model (Holden et al., 2014)</td>
</tr>
<tr>
<td>Regional Climate</td>
<td>ClimgEN</td>
<td>ClimgEN generates regional climate change projections using the method of pattern scaling. (Warren et al., 2012; Osborn et al., 2016)</td>
</tr>
<tr>
<td>Ecosystems, crops, pastures, freshwater</td>
<td>LPjML</td>
<td>Dynamically represents the global terrestrial biosphere (9 natural vegetation types), major crops (12 types), pastures, and optionally bioenergy (two grasses and one tree). Uses ClimgEN projections to simulate coupled carbon, water and vegetation dynamics in response to climate change and human land use. (Kost et al., 2009; Beringer et al., 2011).</td>
</tr>
</tbody>
</table>
3.3. Example 3. PLASIM-ENTSem_ClimGEN_LPJmLem_GEMINI-E3

This coupling (Fig. 3c) is an extension of the PLASIM-ENTSem_GEMINI-E3 one presented in the previous section where the emulator of the agriculture model LPJmLem has been integrated between PLASIM-ENTSem and GEMINI-E3 in order to evaluate physical and economic consequences of climate change on the agricultural sector. For a specified climate policy (see previous section for more details) PLASIM-ENTSem sends climate information at the grid cell level (temperature, precipitation, etc) to LPJmLem that then predicts agricultural variables such as crop yields changes for irrigated or non-irrigated paddy rice, maize and temperate cereal and oil-crop at a spatial resolution of 0.5×0.5°. This information is converted into GEMINI-E3 regions using a conversion key to aggregate the data regionally, and then used to analyse the economic impacts of the selected policy or RCP.

3.4. Example 4: MAGICC_ClimGEN_LPJmLem(crop)_REMIND

This coupling (Fig. 4), which is implemented off-line, uses MAGICC to simulate radiative forcing pathways, global mean temperature and CO₂ time series for the 21st century. ClimGEN generates the corresponding 0.5° regular climate change pattern grid for a selected GCM (eg GFDL-CM2.0). These data are then used to perform climate change impact simulations with the LPJmL bio- and agrosphere model (or alternatively its emulated version LPJmLem, see example 1), focused on variables relevant for use as boundary conditions in the subsequent model chain. LPJmL is set up to provide biophysical inputs to the MAgPIE agro-economy and land use allocation model.
MAgPIE considers the following biophysical constraints on land use patterns, per 0.5° grid cell globally (from LPJmL): (i) Changes in freshwater resources, defined as changes in runoff from the surface and from below-ground and water availability in rivers, lakes and reservoirs; (ii) Changes in soil and vegetation carbon pools; (iii) Changes in potential crop yields of 12 rainfed and irrigated crop types with pasture parameterized in LPJmL, each determined under condition of 7 different management options; (iv) Changes in net irrigation water demand; (v) Sowing and harvest date for all irrigated and rainfed crops.

The first and second of these constraints are examined for potential natural vegetation and the others for both natural and agricultural vegetation. All simulations for the relevant constraints iii-v above were performed for all 7 management options that can be interpreted as different cropping intensities. All runs were made for the two RCPs and the GFDL-CM2.0 GCM and – to separate fertilization and increased water use efficiencies due to enhanced atmospheric CO₂ concentrations – variants were computed in which ambient CO₂ concentration was held constant after year 2002. All results are only used to estimate potentials (irrespective of current land use patterns and management practices) as needed for biophysical constraints in the MAgPIE model. Crop distribution is then calculated by MAgPIE based on the simulated local biophysical potentials.

To determine crop production and land allocation, MAgPIE relies on additional information on bioenergy demand from REMIND (Popp et al., 2011). REMIND computes the bioenergy demand based on a biomass supply curve that uses MAgPIE results from a large number of previous model runs (Klein et al., 2014). In return, MAgPIE gets from REMIND data on greenhouse gas prices. In the RCP3PD scenarios which imply the presence of climate policies, GHG prices represent information on external costs of GHG emitting activities and the urgency of emissions reduction, respectively. Bioenergy is part of a broader technology portfolio that REMIND uses in order to meet the economies’ demand on final energy such as transport energy, electricity, and non-electric energy for stationary end uses. Techno-economic parameters (investment costs, operation & maintenance costs, fuel costs, conversion efficiency etc.) characterize each conversion technology. They essentially determine future technology choice and energy mix. Major outputs from REMIND include primary energy consumption, CO₂ emissions, fossil fuel prices, carbon prices and mitigation costs (i.e. GDP and consumption losses).

3.5. Example 5: PLASIM-ENTSem_TIAMWorld

The objective of the coupling (Fig. 5) of TIAM-WORLD and the emulator of PLASIM-ENTSem is to use regional and seasonal temperature changes obtained from PLASIM-ENTSem in order to represent the possible heating and cooling adjustments due to climate change. Indeed, the climate module included in TIAM-WORLD provides only the global average surface temperature increase. In essence, there is an iterative exchange of data between the two models, whereby TIAM-WORLD sends to the climate emulator a set of total greenhouse gas concentrations for the entire 21st century, computed in TIAM-WORLD, and the climate emulator sends to TIAM-WORLD the seasonal and regional temperatures, converted into seasonal heating and cooling degree-days (HDD/CDD) for each of the regions of the model. PLASIM-ENTSem emulated outputs (seasonal mean and variance of temperature at 5-degree resolution) were converted to heating and cooling Degree days under the assumption that daily temperatures are scattered about the seasonal mean with a normal distribution. These data were integrated onto the 16 TIAM-WORLD regions as a population-weighted average.

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Fig. 3b. The exchange of variables between models in the bespoke framework generator (BFG2) for the PLASIM-ENTSem, GEMINI-E3 coupling. This diagram is generated automatically from the BFG2. The circles denote transformations, the rectangles denote models and the parallelogram denotes outputs from the model. Arrows indicate the direction of data flow within the coupling. Model names within these figures are slightly different from the text as they are program names for example: gemini_e3 is GEMINI-E3.

Fig. 3c. The exchange of variables between models in the bespoke framework generator (BFG2) for the PLASIM-ENTSem, ClimGEN, LPJmL, GEMINI-E3 coupling. This diagram is generated automatically from the BFG2. The rectangles denote models and the parallelogram denotes outputs from the model. Arrows indicate the direction of data flow within the coupling.

Fig. 4. The exchange of variables between models in the bespoke framework generator (BFG2) for the MAGICC, ClimGEN, LPJmL, MAgPIE, REMIND coupling. This diagram is generated automatically from the BFG2. The rectangles denote models and the parallelogram denotes outputs from the model. Arrows indicate the direction of data flow within the coupling.
average temperature rise to 2 °C above pre-industrial levels, although we do not in this study explore scenarios which reduce temperatures more than this.

Early international assessments, such as the IPCC Special Report on Emissions Scenarios (SRES) (Nakicenovic and Swart, 2000) used self-consistent socio-economic scenarios (characterised by population, GDP, land use and energy use) and greenhouse gas emission pathways over time. SRES scenarios were based upon an analysis of how demographic, social, economic, environmental and technological aspects of our society might evolve globally. In these scenarios, two main ‘axes’ of change were considered: (a) environmental versus economic and (b) globalisation versus regionalisation of markets and cultures. Hence, the four scenarios may be briefly summarised as A1 (Global, economic); A2 (Regional, economic); B1 (Global, environmental); B2 (Regional, environmental). A new process, independent of the original SRES scenarios, has since been established (Moss et al., 2010). This recognises that different socioeconomic pathways might have the same climatic change outcome. Hence, SRES scenarios have now been ‘replaced’ by the Representative Concentration Pathways (RCPs), which were used in the IPCC Fifth Assessment Report (AR5) and new Shared Socio-economic Pathways (SSPs) (van Vuuren et al., 2011a; Kriegler et al., 2012; Ebi et al., 2014). In SSPs, the ‘axes’ of change are (a) challenge to mitigation and (b) challenge to adaptation. For example, increased population is a challenge to mitigation because energy demand will be higher. SSPs are based on new set of socio-economic data, including some trends important in SRES such as population and GDP. However, other data may also be important, but most fundamentally, there is a change in the way in which the data are used. The RCPs and SSPs have not been designed as a new, fully integrated and self-consistent set of socio-economic and emission scenarios over time, but instead offer the potential to mix and match alternative combinations. This is undertaken in a framework (a matrix) that combines climate forcing on one axis (as represented by the Representative Forcing Pathways) and socio-economic conditions (represented by the Socio-Economic Pathways) on the other. Thus we apply this new methodology in our research.

Firstly, we ensured that our model couplings were reasonably harmonised in projecting greenhouse gas emissions associated with the RCP6 pathway and the RCP2.6 pathway (Fig. 6). We used the five couplings above (and others) to derive policy relevant information.

The paths of the emissions from the three models GEMINI-E3, REMIND and TIAM-WORLD are illustrated in Fig. 6 alongside reference RCP2.6 and RCP6 trajectories from van Vuuren et al. (2011a) labelled RCP6V and RCP6V2 showing that our simulations from all three models are broadly consistent with theirs. Substantial emissions reductions are needed in order to stabilize the greenhouse gas concentrations in the atmosphere to a level of around 450 ppm CO2eq (RCP2.6). Model couplings including those listed above were used to explore this transition, and to create different strategies for, or implications of reaching these emission reductions. Carbon prices, policy design, energy technologies, and climate change impacts were all
explored. Applying the coupling in example 3, the GEMINI-E3 model explores how the carbon price required to achieve the transition depends on the time of onset of climate change mitigation and on burden sharing approaches to climate change policy. Applying the coupling in example 4, REMIND simulates the most cost-effective way to achieve the emission reductions globally, exploring how this changes when the availability of biomass is low. Other couplings are used to explore the consequences of these emission reductions. Applying the coupling in example 5, TIAM-WORLD simulates the consequences of climate change for heating and cooling demand in the two RCP scenarios, and finally the coupling in example 2 assesses sea level rise impacts in the two scenarios.

Referring to the coupling in example 3, here the GEMINI-E3 model can be used to explore a set of standard burden sharing approaches (Babonneau et al., 2016, and alternative constraints on the date at which climate policy (in the form of the use of a carbon price in international markets) is instigated. Table 3 highlights the key findings. The model indicates that large rises in carbon prices are needed to achieve the necessary emission reduction; however these are greatly reduced if policy is instigated in 2020 rather than 2030. The importance of early policy action has also been highlighted in other studies which report on the implications of short-term emission targets for the cost and feasibility of long-term climate goals such as the 2C target for limiting warming (Luderer et al., 2015; Riahi et al., 2015; Rogelj et al., 2012).

In assessing the transition from RCP6 to RCP2.6, REMIND selects from a large set of potential energy conversion technologies. Generating negative emissions by using biomass in combination with carbon capturing and sequestration turns out to be a favourable, cost-effective option (Fig. 7, left panels). The associated carbon price increases from almost 10 $/tCO2 in 2010 to around 220 $/tCO2 in 2050. The meta-analysis of recent mitigation studies of Clarke et al., (2014) identifies a number of studies that demonstrate feasibility of RCP2.6, whilst emphasizing that higher carbon prices and reliance on bioenergy with carbon capture and storage are necessary to achieve this (Azar et al., 2010). Hence, our results are in line with the findings of many other studies. However, we also explore the effects of limiting the supply of bioenergy (Leimbach et al., 2016). Within both RCP2.6 scenarios (low and high biomass potential) there is a fast phase-out of the coal technologies which are the most carbon-intensive (Fig. 7, upper panels). Importantly, while bioenergy and solar are similarly important for the long-term energy mix in the RCP2.6 scenario (high biomass potential), solar energy is the dominant source of energy in the RCP2.6_biolow scenario. The high sensitivity of the energy system to the availability of biomass can also be seen in Fig. 7 (lower panel), which shows the structure of biomass consumption. In the case of sufficient availability of bioenergy, it is cost effective to produce biofuels for the transport sector. However, it is most cost effective to use biomass to produce hydrogen when the biomass potential is low, as this technology has comparatively lower emissions. Furthermore, hydrogen has the potential to replace fossil resources in sectors other than transport. Coupling outputs suggested that carbon prices up to 600 $/t CO2 were needed to achieve the transition to RCP3PD1 if biofuel cropping was minimised in order to reduce competition for land with agricultural crops and preserve natural ecosystems and biodiversity.

Our studies project that reliance on biofuels for mitigation would induce widespread deforestation and other land use change globally (consistent with the findings of many other studies, e.g. Fargione et al., 2008; Searchinger et al., 2008, Popp et al., 2012, Oppenheimer et al., 2014), unless a carbon taxation scheme is used that includes terrestrial carbon (consistent with the findings of Wise et al., 2009). Our results indicate that the main response option in land-use to climate change mitigation policy is agricultural intensification through investments in yield-increasing technological change. These are estimated to be 41%–72% higher in the policy (RCP3PD) scenario compared to the BAU (business as usual, RCP6) scenario over the 1995 technology level. These are shown in Fig. 8. The role of agricultural intensification has also been highlighted elsewhere (Lotze-Campen et al., 2010; Tilman et al., 2011; Smith et al., 2013).

Results obtained with the coupling PLASIM-ENTSem TIAM-WORLD coupling (example 5) explore the feedback between the climate system and the energy system (Fig. 9). They show that the climate feedback induced by adaptation of the energy system to heating and cooling is found to be insignificant, partly because heating and cooling-induced changes compensate and partly because they represent a limited share of total final energy consumption. However, significant changes are observed at regional levels in the reference case RCP6 (Labriet et al., 2013a,b). In contrast, they are negligible in RCP2.6, with smaller temperature changes. While the increase in cooling demand is met with electricity, the decrease of heating demand results mostly in a decrease

### Table 3

<table>
<thead>
<tr>
<th></th>
<th>2020</th>
<th>2030</th>
<th>2040</th>
<th>2050</th>
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<td>Egalitarian, Slow</td>
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<td>1335</td>
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<tr>
<td>Equalization of cost, Fast</td>
<td>18</td>
<td>63</td>
<td>161</td>
<td>360</td>
</tr>
</tbody>
</table>

![Fig. 7](image1.png)

**Fig. 7.** Upper left: Primary energy consumption in RCP2.6 scenario; upper right: Primary energy consumption in RCP2.6-biolow scenario; lower left: Biomass consumption in RCP2.6 scenario; lower right: Biomass consumption in RCP2.6-biolow scenario; Source REMIND.

![Fig. 8](image2.png)

**Fig. 8.** Yield increases with respect to 1995 due to technological change: Difference between the RCP3PD and the RCP6 scenario for the different assumptions on CO2 fertilization and bioenergy potentials.

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1 RCP3PD corresponds to RCP2.6 and features a peak and decline in radiative forcing.
in gas consumption, this reflects the relatively higher costs of natural gas compared to other energy sources for heating in the longer term. The need for power capacity to satisfy additional cooling services and the pressure on electricity demand result in increases in electricity prices (for example, up to 30% in Europe in the mid-term, and 50% in the long term). Thus climate change was projected to have minimal effects on heating and cooling demand globally, but effects were important regionally, especially in Europe.

A coupled PLASIM-ENTS/GEMINI-E3 sequence is also used to analyse the impacts of sea-level rise (SLR) in the twenty first century. To estimate SLR, we first use the emulator of the climate model PLASIM-ENTS to compute the warming profile related to the GEMINI-E3 baseline scenario. The temperature increase is used to derive SLR using a semi-empirical relationship. Then the physical consequences of SLR are computed using GIS analysis which are incorporated in GEMINI-E3 (see Joshi et al. (2016)). The simulation results suggest that the potential development of future coastal areas is a greater source of uncertainty than the parameters of SLR itself in terms of the economic consequences of SLR. At global level, the economic impact of SLR could be significant when loss of productive land along with loss of capital and forced displacement of populations are considered. Furthermore, highly urbanised and densely populated coastal areas of South East Asia, Australia and New Zealand are likely to suffer significantly if no protective measures are taken. Hence, it is suggested that coastal areas needs to be protected to ameliorate the overall welfare cost across various regions.

Coupled economic and climate models were also exploited in a game theoretical framework to analyse fairness and robustness of the international environmental agreements. First, we identify a total emission budget over the 2010–2050 period that is compatible with the warming at the end of the century being less than 2 °C, according to our climate models. First results show that an acceptable voluntary burden sharing agreement could be obtained among all groups of countries with a balanced welfare loss below 1% of total discounted household consumption. In such an agreement (see Fig. 10), 15.3% of the total emission budget of 424GtC is allocated to USA, 8% to EU, 22.5% to China, 7.5% to India, 4.8% to Russia. In a “robust” solution that prevents potential emissions overshooting in such commitments and takes potential errors arising in the various approximations made in our methodology into consideration, the welfare loss rises to 1.8% for each group of countries. This analysis has recently been extended (see Haurie et al., 2015 and Babonneau et al. 2016).

5. Conclusions

A set of coupled models has been developed within an integrated
framework that can be used in future research projects involving policy makers and other stakeholders, based on the Community Integrated Assessment System, the Bespoke Framework Generator, and the use of statistical emulators for model coupling. We use it to show that when using carbon price mechanisms to induce a transition from a high carbon to a low carbon economy, prices can be minimised if policy action is taken early, if burden sharing regimes are used, and if agriculture is intensified. This is of particular relevance owing to the recent adoption of the Paris Agreement (UNFCCC, 2015). The approach has created long-lasting coupled models available for future policy relevant research. Exploration of the robustness of coupled model outputs to uncertainties should form a key part of this future work.

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Software availability table

Software name BFG2
Developer Rupert Ford
Contact address and postcode STFC Daresbury Laboratory, Warrington WA4 4AD, U.K. Tel.: +44 1925 60 3217 E-mail: rupert.ford@stfc.ac.uk <mailto:rupert.ford@stfc.ac.uk>
Year first available 2005
Hardware required None specific
Software required Python2, libxml2, libxlst, Python lxml2
Program language Python, xslt
Program size Approx. 500 KB (compressed tar file)
Software name SoftIAM
Developers Sudipta Goswami, Santiago de la Nava Santos, Rachel Warren
Contact address Tyndall Centre (contact: Rachel Warren), Tel.: +44 1603 593912; fax: +44 1603 593901. Email: r.warren@uea.ac.uk
Year first available 2004
Hardware required None specific
Software required BFG2
Program language Python
Program size Approx. 370 MB
Availability Model description downloadable from http://www.pik-potsdam.de/research/sustainable-solutions/models/remind
Cost No commercial use
Software name LPJmL vs. 3.2
Developer PIK (LPJmL team) and collaborators
Contact address and postcode Dieter Gerten (responsible scientist), see above; Tel.: +49 331 288 2577; fax: +49 331 288 2695; E-mail: gerten@pik-potsdam.de
Year first available 2007
Hardware required None specific (Unix preferably)
Software required None specific
Program language C
Program size Approx. 1.5MB (tar file, source code only)
Availability Via http://www.pik-potsdam.de/research/projects/cooperations/lpjml or via contact email to enquire for availability
Cost Free for non-commercial use in case of cooperation agreement


