Scenarios towards limiting global mean temperature increase below 1.5 °C

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The 2015 Paris Agreement calls for countries to pursue efforts to limit global-mean temperature rise to 1.5 °C. The transition pathways that can meet such a target have not, however, been extensively explored. Here we describe scenarios that limit end-of-century radiative forcing to 1.9 W m⁻², and consequently restrict median warming in the year 2100 to below 1.5 °C. We use six integrated assessment models and a simple climate model, under different socio-economic, technological and resource assumptions from five Shared Socio-economic Pathways (SSPs). Some, but not all, SSPs are amenable to pathways to 1.5 °C. Successful 1.9 W m⁻² scenarios are characterized by a rapid shift away from traditional fossil-fuel use towards large-scale low-carbon energy supplies, reduced energy use, and carbon-dioxide removal. However, 1.9 W m⁻² scenarios could not be achieved in several models under SSPs with strong inequalities, high baseline fossil-fuel use, or scattered short-term climate policy. Further research can help policy-makers to understand the real-world implications of these scenarios.

cenarios of the energy-economy-land system can facilitate the integrated assessment of the impacts and mitigation of climate change. For the Fifth Assessment Report (AR5) of the Intergovernmental Panel on Climate Change (IPCC), four Representative Concentration Pathways¹ (RCPs) have provided climate researchers with a set of consistent climate forcings²⁻⁴. More recently, the Shared Socio-economic Pathways (SSPs) have been developed^{5,6}. SSPs provide a socio-economic dimension to the integrative work started by the RCPs7. This framework provides a basis of internally consistent socio-economic assumptions that represent development along five distinct storylines8: development under a green-growth paradigm⁹ (SSP1); a middle-of-the-road development along historical patterns¹⁰ (SSP2); a regionally heterogeneous development¹¹ (SSP3); a development that results in both geographical and social inequalities¹² (SSP4); and a development path that is dominated by high energy demand supplied by extensive fossil-fuel use¹³ (SSP5).

Prior to 2015, international climate policy under the United Nations Framework Convention on Climate Change focused on the goal of keeping the global-mean temperature increase below 2° C relative to pre-industrial levels¹⁴. The Paris Agreement reset this long-term goal to holding the increase well below 2° C and pursuing efforts to limit it to 1.5° C¹⁵. In this study, we present a set of stringent climate change mitigation scenarios consistent with an increase of 1.5° C in 2100. Six integrated assessment models were included in this study (AIM, the Asia–pacific Integrated Model¹¹; GCAM4, the Global Change Assessment Model¹²; IMAGE, the Integrated

Model to Assess the Global Environment⁹; MESSAGE-GLOBIOM, the Model for Energy Supply Strategy Alternatives and their GeneralEnvironmental Impact combined with the Global Biosphere Management Model¹⁰; REMIND-MAgPIE, the Regionalized Model of Investments and Development combined with the Model of Agricultural Production and its Impact on the Environment^{13;} and WITCH-GLOBIOM, the World Induced Technical Change Hybrid model combined with GLOBIOM¹⁶), with which we attempted to model scenarios that limit end-of-century radiative forcing to 1.9W m⁻² under various SSPs (hereafter called 'SSPx-1.9' scenarios, with SSPx indicating the specific SSP assumed by the scenario and 1.9 the radiative forcing target in 2100, Methods). This scenario set allows the structured exploration of climate change at a level consistent with limiting the global-mean temperature increase in 2100 to 1.5 °C with approximately 66% probability (see Fig. 1 and results described below). Overall, all teams were able to produce 1.9W m⁻² scenarios in SSP1, and four teams were successful in SSP2. Of the three and four modelling frameworks that attempted to model 1.9 W m⁻² scenarios in SSP4 and SSP5, one and two were successful, respectively (see Methods, Supplementary Table 1, Supplementary Fig. 1, Supplementary Text 2). From this set of 1.9 W m⁻² scenarios, a further, stringent climate mitigation scenario has been selected for inclusion in the Scenario Model Intercomparison Project¹⁷ (ScenarioMIP) of the Sixth Phase of the Coupled Model Intercomparison Project¹⁸ (CMIP6), as well as other CMIP6 MIPs (for example, refs 19,20, Fig. 1a, Supplementary Text 1, Methods).

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Fig. 1 | Emission and temperature characteristics of 1.9 W m⁻² **scenarios under various SSPs. a**, Global CO₂ emissions of SSP scenarios with the selected CMIP6 ScenarioMIP subset highlighted. Historical emission from ref. ⁵². All other panels show 1.9 W m⁻² scenario data only. **b**, Global Kyoto GHG emissions. Shaded areas show the range per SSP, solid lines the marker scenarios for each SSP and dashed lines single scenarios that are not markers. Single model detail is provided in Supplementary Fig. 2. **c**, Non-CO₂ GHGs per scenario in 2100. **d**, Exceedance probability of various temperature limits for the 1.9 W m⁻² scenarios with bars showing the full range over all available scenarios per SSP. Except for the first sub-panel, all other panels give the exceedance probability over the entire twenty-first century. e, Probability of peak warming versus 2030 GHG emissions in 1.9 W m⁻² scenarios. **f**, Dependence of cumulative CO₂ emissions on non-CO₂ radiative forcing in 2100.

Emission and climate-related outcomes

 CO_2 and other greenhouse gas (GHG) emissions peak before 2030 and decline rapidly over the next two to three decades in SSPx-1.9 scenarios (Fig. 1, see Supplementary Figs. 2–6 for other emissions). By 2050, annual CO_2 and GHG emissions are in the range of -9-6 and 1–13 billion tons of CO_2 -equivalent emissions (gigaton Gt CO_2 e yr⁻¹, Methods), respectively, across all available scenarios. Underlying these reductions is a phase-out of industry and energyrelated CO_2 production at a rate of 0.2–7.1% yr⁻¹ (median: 3.0% yr⁻¹, see Supplementary Tables 2, 3 for a complete overview), combined with rapid upscaling of carbon capture and storage (CCS) and carbon-dioxide removal (CDR, see section on system transformations below). Near-term emissions vary across the SSPs, because, in contrast to SSP1, the effectiveness of near-term climate policies is assumed to be limited in other SSPs (defined by so-called Shared Policy Assumptions^{5,21}). In that case, global mitigation is regionally scattered and accelerates slower over the next few decades, requiring it to accelerate faster later on.

All scenarios presented here lead to 1.9 W m^{-2} radiative forcing in 2100 within rounding precision (Supplementary Fig. 7), but they differ in their likelihood of limiting warming below specific temperature levels. All scenarios keep warming to below 2 °C with more than 66% probability (Fig. 1d), and maximum (peak) median temperature estimates vary from 1.5 °C to 1.8 °C. Near-term mitigation has a determining role here: higher 2030 emissions come with a temperature penalty (Supplementary Fig. 8). The probability of limiting peak warming to below 1.5 °C relative to pre-industrial levels is approximately halved and peak temperature about 0.2 °C higher if emissions are at the high (>45 GtCO₂e yr⁻¹) instead of the low (<30 GtCO₂e yr⁻¹) end of the available range in 2030 (Fig. 1e). By 2100, this variation disappears and all scenarios limit warming below 1.5 °C with about 66% probability (Supplementary Figs. 8, 9).

Whether these pathways provide an acceptable interpretation of the Paris Agreement long-term temperature goal is not a scientific but a political question^{22,23}, which we do not address here.

Across all 13 available scenarios, net zero GHG emissions are reached around 2055-2075 (rounded to the nearest 5 years). Net zero CO₂ emissions are reached earlier (Supplementary Table 2). The year of reaching net zero GHG emissions is inversely correlated with emissions in 2030. For example, scenarios with 2030 GHG emissions higher than 40 GtCO₂e yr⁻¹ reach global net zero GHG emissions before 2060 (Supplementary Fig. 10). Cumulative CO₂ emissions over the 2016–2100 period range from –175 to 475 GtCO₂ (SSP2 median: 250 GtCO₂, rounded to the nearest 25 GtCO₂). Endof-century non-CO₂ radiative forcing strongly influences the variation across this range²⁴ (Fig. 1f). These values are consistent with earlier published estimates²⁴⁻²⁷ (Supplementary Text 3) and lead to 2100 atmospheric CO₂ concentrations in the 350-390 p.p.m. range. Potential feedbacks that are currently not included, such as CO₂ and CH₄ release from permafrost thawing or changes in other natural sources, can reduce carbon budgets further^{28,29} and therefore alter the presented climate outcomes.

Even in these very stringent mitigation pathways, sizeable remaining CH_4 and N_2O emissions are projected by all models (Fig. 1c, Supplementary Fig. 6), and in 2100, respectively, 53–85% and 59–95% of these emissions originate from agriculture. The uncertainty in CH_4 and N_2O emissions is large with inter-model variations dominating inter-SSP variations. High and low estimates for 2100 differ by a factor of 2–3, mainly owing to uncertainties in how emissions from agriculture are treated and can be mitigated in different models^{30,31}. Important uncertainties also remain in the CO_2 mitigation contribution of the land-use sector³¹ (Supplementary Fig. 5). Here, emissions decline over the long term, but whether and to what degree the land-use sector becomes a global sink is very model-dependent (Supplementary Text 4).

System transformations

Achieving pronounced emission reductions requires a transformation of the global economy. Previous studies have discussed the implications of such a global transformation for the energy and land-use system³², highlighting the importance of limiting future energy demand³² to keep warming to below 1.5 °C and of changing consumption patterns³³ combined with sustainable intensification of agriculture³⁴. We here focus on confirming these characteristics and exploring the extent to which they vary across SSPs.

All 1.9 W m⁻² scenarios in this study strongly limit energy demand growth (Fig. 2d, Supplementary Fig. 11), with energy intensity reduction rates of 2-4% yr⁻¹ from 2020 to 2050 (Fig. 2d). In SSP2, final energy demand in 2050 is limited to 10–40% above 2010 levels (rounded to the nearest 5%). This compares to 10% below to 30% above, and 45–75% above 2010 levels in SSP1 and SSP5, respectively. Energy conservation is therefore a common strategy in stringent mitigation scenarios, but it also has limits.

Energy supply also has to be transformed to achieve reductions in deep emissions. This includes upscaling of bioenergy and renewable energy technologies, shifting away from freely emitting fossil-fuel use, and the deployment of CDR, such as Bioenergy with Carbon Capture and Sequestration (BECCS) or large-scale afforestation (see Supplementary Text 5 for a discussion of CDR in SSPx-1.9 scenarios). Non-biomass renewables (solar, wind, hydro and geothermal energy) scale up rapidly over the twenty-first century (Fig. 2a), reaching mid-century electricity shares of 60–80% and 32–79% in SSP1 and SSP2, respectively (Supplementary Fig. 12). In the marker SSP scenarios, these shares are 79%, 60% and 61% in SSP1, SSP2 and SSP5, respectively. Both solar and wind energy is projected to scale up consistently across the different SSPs (Supplementary Fig. 13). Particularly for wind energy, inter-model variations dominate over differences induced by different SSPs, a feature also present in less

stringent mitigation pathways³⁵ (Supplementary Table 4). SSP2 and SSP5 1.9 W m⁻² scenarios see a strong upscaling of nuclear power, whereas in SSP1, and particularly its marker implementation, the contribution of nuclear energy use decreases compared to today's levels (Supplementary Fig. 13).

Under all SSPs, $1.9 \text{ W} \text{ m}^{-2}$ scenarios show a clear shift away from unabated fossil fuels (that is, without CCS, Fig. 2c), and a phaseout of all fossil fuels. The marker implementations exhibit rapidly declining contributions of coal until 2040 (less than about 20% of its 2010 contribution in 2040), followed by a phase-out of oil until 2060 (Supplementary Figs. 14, 15). The potential contribution of natural gas to the primary energy mix is the most uncertain, with mid-century contributions ranging from 22 to 267 exajoules (EJ) yr⁻¹ across all scenarios compared to about 100–110 EJ yr⁻¹ in 2010. Differences in preferences for gas supply across models here dominate the variation in costs and availability assumptions owing to alternative socio-economic pathways (Supplementary Table 4, Supplementary Fig. 16).

Bioenergy is used in large amounts in all 1.9 W m⁻² scenarios, and this can raise concerns for food security or biodiversity³⁶⁻³⁸. These concerns depend both on how and how much bioenergy is produced. Bioenergy demands can be met through dedicated energy crops or through residues. The latter option comes with fewer tradeoffs than dedicated bioenergy crops³⁸. Models, however, project very different shares for the use of residues (Supplementary Table 5), and further research clarifying its potential would be essential. For 2050, global technical bioenergy potentials (including energy crops and residues) were identified ranging from <50 to >500 EJ yr⁻¹. High, medium and low agreement was attributed to potentials of 100, 300 and $>300 \text{ EJ yr}^{-1}$, respectively³⁶. Bioenergy use is increased by 1–5% per year between 2020 and 2050 in $1.9\,\mathrm{W}\,\mathrm{m}^{-2}$ scenarios. Total bioenergy use in 2050 is kept below about 300 EJ yr⁻¹, and in most cases below 150 EJ yr⁻¹ (Supplementary Fig. 17). In a green-growth SSP1 world, markedly lower bioenergy contributions are projected compared to an SSP2 world that continues the historical experience (34-112 EJ yr⁻¹ lower in 2050). Putting this into context, scenarios project approximately 100 EJ yr⁻¹ of bioenergy use (full range: 38-112, with important variations across SSPs) in baseline scenarios without any climate policy (Supplementary Fig. 17).

In 1.9 W m⁻² scenarios, land for energy crops and forest area is generally projected to expand during the twenty-first century, with large variations across models, and this can impact land for agriculture and water availability^{39,40} (Fig. 2f, Supplementary Fig. 18). However, in SSP1 the decrease in agricultural land in 1.9W m⁻² scenarios is reasonably similar to what is projected in a no-climate-policy baseline merely owing to low demand for agricultural commodities and high agricultural intensification. Pasture is one of the activities most affected by expanding other land uses and declines robustly across models and SSPs (Supplementary Fig. 19). In the middle-of-the-road SSP2 world, pastures decreases by 1-20% in 2050 compared to 2010 levels, and in SSP1, pastures also decrease by 8-16%. In a fossil-fuel intensive SSP5 scenario, it declines by 15-25%. It is important to note that SSP1 baseline scenarios already project a pasture-land decrease of 1–11% due to shifts towards less meat-intensive diets, limited food waste and a return of the world population to 7 billion people by 2100^{5,9,31}. This reaffirms the important role that changes in food consumption in combination with sustainable intensification of agriculture have for stringent mitigation^{31,34,41}.

Large-scale afforestation and reforestation can make an important contribution to the overall CDR effort. In the sustainable SSP1 world, pressure on land is relatively low, and the forest area in 2050 can therefore expand by 0-24% relative to 2010. However, in the middle-of-the-road SSP2 scenarios, results are mixed, with some models projecting forest area to decrease by 2% and others report an increase of up to 18%. SSP5 sees a change of 0-16%(Supplementary Table 6). Not all models explicitly include afforestation as a mitigation option and ranges therefore span results that

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Fig. 2 | Overview of key decarbonization characteristics in 1.9 W m⁻² scenarios. **a**, Primary energy from non-biomass renewables (wind, solar, hydro and geothermal energy). **b**, Primary energy from biomass with CCS (BECCS). **c**, Primary energy from coal without CCS. Shaded areas in **a-c** show the range per SSP, solid lines the marker scenarios for each SSP and dashed lines single scenarios that are not markers. **d**, Three illustrations of global final energy demand in 1.9 W m⁻² scenarios showing, from left to right, the average reduction from baseline over the 2020-2100 period, the change in 2050 compared to 2010 levels, and the annual rate of final energy intensity change. **e**, Global forest cover, and change relative to 2010 due to afforestation and reforestation in 2.6 and 1.9 W m⁻² scenarios. **f**, Change in global cropland for agriculture in 2100 relative to 2010 in 'Baseline' scenarios in the absence of climate change mitigation, as well as in 2.6 and 1.9 W m⁻² scenarios. Results are grouped per SSP (coloured lines with black symbols). rel., relative. Mha, million hectares.

are not fully comparable across models. However, in all 1.9 W m⁻² scenarios climate policy leads to a net forest expansion compared to no-climate-policy baselines (Fig. 2e). Integrated policy packages are required that ensure food security is achieved together with climate change mitigation⁴².

BECCS contributes the largest part of CDR in 1.9 W m^{-2} scenarios (Supplementary Fig. 20). Between $150-1,200 \text{ GtCO}_2$ (rounded to nearest 25 GtCO₂), equivalent to about 4–30 years of current annual emissions, is removed from the atmosphere via BECCS during the twenty-first century, with important variation between models and across SSPs (Fig. 3a, d). SSP1 shows the lowest BECCS deployment over the twenty-first century ($150-700 \text{ GtCO}_2$) owing to its lower final energy demand and baseline emissions, compared to SSP2 ($400-975 \text{ GtCO}_2$) and SSP5 ($950-1,200 \text{ GtCO}_2$). None of the SSPx-1.9 scenarios explicitly attempted to limit the contribution from BECCS. The numbers reported here therefore represent projections of estimated cost-effective BECCS deployment in 1.9 W m⁻² scenarios, but do not represent minimum BECCS requirements in a strict sense.

Abated fossil fuels—that is, fossil fuels combined with CCS (fossil–CCS)—are often used in models as a bridging solution. However, fossil–CCS still results in residual CH₄ emissions from coal mining or gas handling, and CO₂ emissions due to imperfect capture and leakage. These emissions can become too substantial for very stringent mitigation transitions. Indeed, almost all 1.9 W m⁻² scenarios deploy less cumulative fossil–CCS than weaker mitigation scenarios (Fig. 3c). Optimal 1.9 W m⁻² strategies are therefore not merely 'more of the same'. Overall, the BECCS share of total CCS increases (Supplementary Fig. 20). CDR is thus preferred over fossil–CCS in very stringent mitigation scenarios.

Differential mitigation

A previous study⁴³ has identified characteristics of 1.5 °C pathways in comparison to 2°C pathways. These characteristics were (i) greater mitigation efforts on the demand side; (ii) energy efficiency improvements; (iii) CO₂ reductions beyond global net zero; (iv) additional GHG reductions mainly from CO_2 ; (v) rapid and profound near-term decarbonization of energy supply; (vi) higher mitigation costs; and (vii) comprehensive emission reductions implemented in the coming decade. Using our 1.9 W m⁻² and 2.6 W m⁻² scenarios as proxies for 1.5 °C and 2 °C pathways, these characteristics still hold when assessed with four additional models and varying socio-economic assumptions (Fig. 4, Supplementary Text 6, and results above). None of the 1.9 W m⁻² scenarios show a peak of emissions after 2020, and 82-98% of additional cumulative mitigation over the 2020-2100 period is achieved through CO₂ reductions (Supplementary Fig. 21). Fig. 4 further illustrates the relatively stronger demand-side mitigation efforts in 1.9 W m⁻² scenarios, particularly in the transport and building sectors (see also Supplementary Figs. 22-24).

Mitigation costs increase substantially between 1.9 and 2.6 W m^{-2} scenarios reflecting higher marginal abatement costs (Figs. 4,5). The relative carbon price increase is largest in SSP2 (Fig. 4) and also SSP1 sees large relative increases across all models (Supplementary Figs. 22–24). However, in absolute terms, carbon

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Fig. 3 | BECCS, **fossil-CCS and CCS across SSPs and across climate targets. a**, Annual amount of CO_2 stored by CCS in 1.9 W m⁻² scenarios. Shaded areas show the range per SSP, solid lines the marker scenarios for each SSP and dashed lines single scenarios that are not markers. **b**, Variation per modelling framework and per SSP of cumulative CO_2 stored by CCS during the twenty-first century when moving from a world in the absence of climate policy (baseline) to increasingly more stringent climate targets (6.0, 4.5, 3.4, 2.6 and 1.9 W m⁻²) **c,d**, As **b** but for fossil-CCS and BECCS, respectively. Note that axis limits vary across models.

prices (Fig. 5), consumption losses and energy supply mitigation investments (Supplementary Fig. 26) are highest when assuming the less favourable socio-economic conditions of SSP2, SSP4 and SSP5. For instance, the average discounted carbon prices (discounted to 2010 over the 2020–2100 period; Fig. 5) are estimated to be about 50–165 US\$ per tCO₂e in SSP2 (rounded to the nearest 5). They are approximately 35–65% lower in SSP1, and for the two reported SSP5 scenarios the change is -30% and +5%, respectively. The large range of carbon prices is mainly driven by model uncertainties, which were already identified for 2.6 W m⁻² scenarios⁵, but are more pronounced here owing to the more stringent target.

Enabling and disabling factors

Our results show that some socio-economic developments and assumptions about policy effectiveness preclude achieving stringent mitigation futures (Fig. 5). Such failures were anticipated for SSP3, in which very heterogeneous regional development and debilitating policy assumptions already rendered limiting end-ofcentury radiative forcing to 2.6 W m⁻² unachievable in the models⁵ (Supplementary Text 2). However, in SSP4 and SSP5 limiting radiative forcing to 1.9 W m⁻² proved difficult too. In SSP4, a world that promotes both geographical and social inequalities, only one out of three models attempting a $1.9\,W\,m^{-2}$ scenario was successful. Weak mitigation is achieved rather easily in SSP4^{5,12}. However, the lack of control over land-related emissions in developing countries and lower acceptability of CCS in developed countries in SSP4 make very low emissions pathways unachievable¹². Also in SSP5, a world dominated by high economic growth and fossil-fuel development, challenges to mitigation are high¹³. Finally, under a middle-of-theroad development (SSP2) and under a green-growth paradigm (SSP1) four and six models, respectively, were able to produce a 1.9 W m⁻² scenario (Supplementary Table 1).



Fig. 4 | Differential mitigation characteristics when moving from a 2.6 W m⁻² SSP2 scenario to a 1.9 W m⁻² scenario under three SSP assumptions (SSP1, SSP2, SSP5). Updated from ref. ⁴³. Indicators are: long-term mitigation costs (2010–2100 aggregate consumption losses relative to baseline discounted at 5%); short-term mitigation costs (2010–2040 aggregate discounted at 5%); 2040 global emission-weighted equivalent carbon price level; electricity price in 2030; cumulative CDR between 2010 and 2100 including BECCS and CO₂ removal by land use and land-use change; decarbonization pace (average linear 2010–2050 rate of reductions in energy-related CO₂ emissions); reductions in CO₂ emissions from electricity from baseline in 2050; reductions in CO₂ emission from transport from baseline in 2050; and reductions in CO₂ emissions from buildings from baseline in 2050. Data are shown for the marker implementations of SSP1, SSP2 and SSP5. Ranges per SSP are provided in Supplementary Figs. 22–24. RF, radiative forcing. *Not available for all models.



Fig. 5 | Variation of carbon prices over SSP and radiative forcing target space. Values are shown as average global average carbon prices over the 2020-2100 period discounted to 2010 with a 5% discount rate. Mitigation challenges are assumed to increase from left to right across the SSPs (that is, SSP1, SSP4, SSP2, SSP3, SSP5). Each box represents one model–SSPradiative forcing target combination. A, AIM/CGE; G, GCAM4; I, IMAGE; M, MESSAGE-GLOBIOM; R, REMIND-MAgPIE; W, WITCH-GLOBIOM. All scenarios with a carbon price greater than 0 (that is, all but the baselines) have been designed to reach one of the radiative forcing targets on the vertical axis. Models for which no baseline data are indicated have baselines that result in an end-of-century radiative forcing between 6.0 and 8.5 W m⁻².

Mitigation challenges for achieving a 1.9 W m⁻² target thus differ strongly across the SSPs, as illustrated in Fig. 6. For example, the amount of CO₂ emission that has to be avoided varies by a factor of two between SSP1 and SSP5 worlds in 1.9 W m⁻² scenarios (Fig. 6a). The projected use of BECCS varies by a factor 2 to almost 3 between SSP1, and SSP2 and SSP5, respectively (Fig. 6c), and also land-use CO₂ mitigation contributions vary massively yet less distinctly (Fig. 6b). Furthermore, the shift away from baseline development implied by the energy system transformation is also markedly smaller in SSP1 than in SSP2 or SSP5 (Fig. 6d-f), and therefore comes with potentially lower overall societal hurdles. Even when overcoming these differences in starting points, the difficulty or facility of achieving deep mitigation remains very diverse across SSPs. In particular, the lower level of final energy demand that can be achieved in SSP1 implies a smaller energy supply system^{5,35} (Fig. 6g) and thus also a smaller amount of investment needs to decarbonize it (Fig. 6h). Finally, also residual emissions from agriculture and the emission intensity of food production differ strongly between SSPs (Fig. 6i,j) highlighting that challenges have to be overcome in all sectors. Each of these dimensions identifies possibilities for potential policy intervention.

Interpretation and feasibility

What can SSPx-1.9 scenarios teach us about the feasibility of limiting warming to 1.5 °C? Typically, feasibility refers to a multi-dimensional concept that considers aspects of geophysics, technology, economics, societal acceptance, institutions and politics, among other disciplines. In this context, integrated scenarios provide

High mitigation Low mitigation challenges challenges а SSP5 SSP4 SSP2 SSP1 \$77 2.000 4.000 5.000 6,000 7.000 8,000 3.000 b 717 SSP4 SSP2 200 -700 -600 -500 -400 -300 -200 -1000 100 c SSP5 SSP4 SSP2 SSP1 £ 15 10 0 5 d SSP5 SSP4 SSP2 SSP1 2 5 6 8 9 10 11 12 3 4 7 e ☆ 50 100 150 200 250 300 350 f ☆ 50 55 60 65 70 75 80 g ☆ -SSP5 SSP4 SSP2 700 300 350 400 450 500 550 600 650 h SSP5 SSP4 SSP2 SSP1 ☆ 6 5 i SSP5 SSP4 SSP2 SSP1 ☆ 2 3 5 6 8 9 Δ j SSP5 SSP4 SSP2 0.9 0.3 0.4 0.5 0.6 0.7 0.8 SSP2 SSP4 SSP5 Models: X : AIM/CGE (A); + : GCAM4 (G); O : IMAGE (I); SSP marker: □ : REMIND-MAGPIE (R); : MESSAGE-GLOBIOM (M);

☆ : WITCH-GLOBIOM (W)

Fig. 6 | Variation in mitigation challenges for limiting end-of-century radiative forcing to 1.9 W m⁻² across the SSPs. a-j, Various dimensions of climate change mitigation challenges are shown. A description of the ten indicators shown here is provided in Supplementary Table 7. Ranges show the minimummaximum range across models per SSP. Symbols show single models. The yellow line indicates the marker implementation for each respective SSP. As not all modelling frameworks provide all necessary indicators, some panels show fewer models. No model was able to produce a 1.9 W m⁻² scenario for SSP3. a, Cumulative CO₂ emission reduction from baseline in the 2020-2100 period for the 1.9 W m⁻² scenarios (GtCO₂). **b**, Cumulative net land-use CO₂ in the 2020-2100 period for the 1.9 W m⁻² scenarios (GtCO₂). c, Average annual CO₂ storage from BECCS for the 1.9 W m⁻² scenarios for the 2020-2100 period (GtCO₂ yr⁻¹). d, Upscaling of low-carbon primary energy share for the 1.9 W m⁻² scenarios in 2050 relative to baseline. e, Reduction in coal primary energy for the 1.9 W m⁻² scenarios in 2050 relative to baseline (EJ yr⁻¹). f, Reduction in carbon intensity of primary energy in 2050 for the 1.9 W m⁻² scenarios relative to baseline (tCO $_2$ TJ⁻¹). **g**, Average final energy demand for the 2020–2100 period for the 1.9 W m⁻² scenarios (EJ yr⁻¹). **h**, Average annual energy system investment for the 2020–2100 period for the 1.9 W $m^{\text{-2}}$ scenarios (trillion 2005 US\$). \mathbf{i} , Non-CO₂ emissions from agriculture for the 1.9 W m⁻² scenarios in 2050 (GtCO₂e yr⁻¹). j, Emission intensity of food production for the 1.9 W m⁻² scenarios in 2050 (gCO₂e kcal⁻¹).

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insights about the technological and economic assumptions under which a global climate goal could or could not be achieved. However, because models are stylized, imperfect representations of the world, feasible dynamics in a model might be infeasible in the real world, while vice versa infeasibility in a model might not mean that an outcome is infeasible in reality.

For example, modelled energy transition pathways assume broad social acceptance, convergence towards global cooperation, and limited political inertia or institutional barriers—conditions that are different in reality. At the same time, reality can also move faster than assumed in models⁴⁴. Advanced and pervasive information technologies that dominate our lives today would not have been considered feasible half a century ago, and also recent real-world cost reductions for renewable energy technologies exceeded expectations even of the more optimistic scenarios from 20 years ago.

Earlier studies have highlighted the importance of deriving insights from scenarios that are able to reach the intended target, and scenarios that indicate under which conditions a target cannot be met⁴⁵. This has led to the development of more sophisticated interpretations of structured scenario ensembles, which suggest that the proportion of successful scenario results can be used as an indicator of infeasibility risk⁴⁶. In this context, our scenarios can illustrate that multiple technologically salient options are available for limiting warming increase to 1.5 °C, but that the risk of failure increases markedly in the high growth, unequal and/or energyintensive worlds of SSP3, SSP4 and SSP5. Any interpretation of models that are unable to reach a certain target comes with caveats, because models, including IAMs, are coarse approximations of reality. Real-world feasibility of a particular scenario also depends on factors that are not covered by current IAMs (such as social support) or enabling factors (such as rapid technological developments). These might shift assessments of feasibility in either a more positive or negative direction.

The policy scenarios reported here thus inform certain aspects, but should not be considered as an absolute statement on feasibility³². Policy analysts and advisors still need to translate the insights of this and other related studies^{39,43,47–51} into a more complete assessment of feasibility, which accounts for the broader context of societal preferences, politics and recent real-world trends.

Going forward

This study aimed to develop a set of stringent integrated community scenarios that can facilitate the assessment of climate impacts, mitigation and adaptation challenges in the context of the Paris Agreement. However, continued research is needed. A stronger involvement of the social sciences that study how societies change and transform can provide valuable complementary insights. To facilitate such further analyses, data presented here are made available to the wider community. Finally, the SSP1-1.9 marker implementation will be included as a very low climate change scenario in CMIP6 ScenarioMIP (Supplementary Text 1), and detailed climate data for these scenarios will become available during the 2018–2020 time period^{17,18}.

Methods

Methods, including statements of data availability and any associated accession codes and references, are available at https://doi. org/10.1038/s41558-018-0091-3.

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References

- l. van Vuuren, D. et al. The representative concentration pathways: an overview. *Climatic Change* **109**, 5–31 (2011).
- Taylor, K. E., Stouffer, R. J. & Meehl, G. A. An overview of CMIP5 and the experiment Design. *Bull. Am. Meteorol. Soc.* 93, 485–498 (2011).

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- Warszawski, L. et al. The inter-sectoral impact model intercomparison project (ISI-MIP): project framework. *Proc. Natl Acad. Sci. USA* 111, 3228–3232 (2014).
- Meinshausen, M. et al. The RCP greenhouse gas concentrations and their extensions from 1765 to 2300. *Climatic Change* 109, 213–241 (2011).
- Riahi, K. et al. The Shared Socioeconomic Pathways and their energy, land use, and greenhouse gas emissions implications: an overview. *Glob. Environ. Change* 42, 153–168 (2017).
- O'Neill, B. et al. A new scenario framework for climate change research: the concept of shared socioeconomic pathways. *Climatic Change* 122, 387–400 (2014).
- van Vuuren, D. P. et al. A new scenario framework for climate change research: scenario matrix architecture. *Climatic Change* 122, 373–386 (2014).
- O'Neill, B. C. et al. The roads ahead: narratives for shared socioeconomic pathways describing world futures in the 21st century. *Glob. Environ. Change* 42, 169–180 (2017).
- 9. van Vuuren, D. P. et al. Energy, land-use and greenhouse gas emissions trajectories under a green growth paradigm. *Glob. Environ. Change* **42**, 237–250 (2017).
- 10. Fricko, O. et al. The marker quantification of the Shared Socioeconomic Pathway 2: a middle-of-the-road scenario for the 21st century. *Glob. Environ. Change* **42**, 251–267 (2017).
- 11. Fujimori, S. et al. SSP3: AIM implementation of Shared Socioeconomic Pathways. *Glob. Environ. Change* **42**, 268–283 (2017).
- 12. Calvin, K. et al. The SSP4: a world of deepening inequality. *Glob. Environ. Change* **42**, 284–296 (2017).
- Kriegler, E. et al. Fossil-fueled development (SSP5): an energy and resource intensive scenario for the 21st century. *Glob. Environ. Change* 42, 297–315 (2017).
- Decision 1/CP.16 The Cancun Agreements: Outcome of the Work of the Ad Hoc Working Group on Long-term Cooperative Action Under the Convention (UNFCCC, 2010).
- 15. Paris Agreement (UNFCCC, 2015).
- Emmerling, J. et al. The WITCH 2016 model documentation and implementation of the Shared Socioeconomic Pathways. FEEM Working Paper 42.2016 (2016).
- O'Neill, B. C. et al. The scenario model intercomparison project (ScenarioMIP) for CMIP6. *Geosci. Model Dev.* 9, 3461–3482 (2016).
- Eyring, V. et al. Overview of the coupled model intercomparison project phase 6 (CMIP6) experimental design and organization. *Geosci. Model Dev.* 9, 1937–1958 (2016).
- Jones, C. D. et al. C4MIP the coupled climate-carbon cycle model intercomparison project: experimental protocol for CMIP6. *Geosci. Model Dev.* 9, 2853–2880 (2016).
- Lawrence, D. M. et al. The land use model intercomparison project (LUMIP) contribution to CMIP6: rationale and experimental design. *Geosci. Model Dev.* 9, 2973–2998 (2016).
- Kriegler, E. et al. A new scenario framework for climate change research: the concept of shared climate policy assumptions. *Climatic Change* 122, 401–414 (2014).
- 22. Schleussner, C.-F. et al. Science and policy characteristics of the Paris Agreement temperature goal. *Nat. Clim. Change* **6**, 827–835 (2016).
- Knutti, R., Rogelj, J., Sedlacek, J. & Fischer, E. M. A scientific critique of the two-degree climate change target. *Nat. Geosci.* 9, 13–18 (2016).
- Rogelj, J. et al. Differences between carbon budget estimates unravelled. Nat. Clim. Change 6, 245–252 (2016).
- IPCC Climate Change 2014: Synthesis Report (eds Core Writing Team, Pachauri, R. K. & Meyer L. A.) (IPCC, 2015).
- Clarke, L. et al. in *Climate Change 2014: Mitigation of Climate Change* (eds Edenhofer, O. et al.) 413–510 (IPCC, Cambridge Univ. Press, 2014).
- Rogelj, J. et al. Energy system transformations for limiting end-of-century warming to below 1.5°C. Nat. Clim. Change 5, 519–527 (2015).
- MacDougall, A. H., Zickfeld, K., Knutti, R. & Matthews, H. D. Sensitivity of carbon budgets to permafrost carbon feedbacks and non-CO₂ forcings. *Environ. Res. Lett.* 10, 125003 (2015).
- 29. Schneider von Deimling, T. et al. Estimating the near-surface permafrostcarbon feedback on global warming. *Biogeosciences* 9, 649-665 (2012).
- Gernaat, D. E. H. J. et al. Understanding the contribution of non-carbon dioxide gases in deep mitigation scenarios. *Glob. Environ. Change* 33, 142–153 (2015).
- Popp, A. et al. in Land-use futures in the shared socio-economic pathways. Glob. Environ. Change 42, 331–345 (2017).
- Clarke, L. et al. in *Climate Change 2014: Mitigation of Climate Change* (eds Edenhofer, O. et al.) Ch. 6, 413–510 (IPCC, Cambridge Univ. Press, 2014).
- Popp, A., Lotze-Campen, H. & Bodirsky, B. Food consumption, diet shifts and associated non-CO₂ greenhouse gases from agricultural production. *Glob. Environ. Change* 20, 451–462 (2010).
- Havlík, P. et al. Climate change mitigation through livestock system transitions. Proc. Natl Acad. Sci. USA 111, 3709–3714 (2014).

- Bauer, N. et al. Shared Socio-Economic Pathways of the energy sector — quantifying the narratives. *Glob. Environ. Change* 42, 316–330 (2017).
- Creutzig, F. et al. Bioenergy and climate change mitigation: an assessment. Glob. Change Biol. Bioenergy 7, 916–944 (2015).
- Bonsch, M. et al. Trade-offs between land and water requirements for large-scale bioenergy production. *Glob. Change Biol. Bioenergy* 8, 11–24 (2016).
- Smith, P. et al. in *Climate Change 2014: Mitigation of Climate Change* (eds Edenhofer, O. et al.) Ch. 11, 811–922 (IPCC, Cambridge Univ. Press, 2014).
 Smith, P. et al. Biophysical and economic limits to negative CO₂ emissions.
- Nat. Clim. Change 6, 42–50 (2016). 40. Field, C. B. & Mach, K. J. Rightsizing carbon dioxide removal. Science 356,
- 706-707 (2017).
 41. Smith, P. et al. How much land-based greenhouse gas mitigation can be achieved without compromising food security and environmental goals? *Glob. Change Biol.* 19, 2285-2302 (2013).
- Valin, H. et al. Agricultural productivity and greenhouse gas emissions: trade-offs or synergies between mitigation and food security? *Environ. Res. Lett.* 8, 035019 (2013).
- Rogelj, J. et al. Energy system transformations for limiting end-of-century warming to below 1.5 °C. Nat. Clim. Change 5, 519–527 (2015).
- 44. Creutzig, F. et al. The underestimated potential of solar energy to mitigate climate change. *Nat. Energy* **2**, 17140 (2017).
- Tavoni, M. & Tol, R. Counting only the hits? The risk of underestimating the costs of stringent climate policy. *Climatic Change* 100, 769–778 (2010).
- Riahi, K. et al. Locked into Copenhagen pledges implications of short-term emission targets for the cost and feasibility of long-term climate goals. *Technol. Forecast. Soc. Change* **90**, 8–23 (2015).
- Sanderson, B. M., O'Neill, B. C. & Tebaldi, C. What would it take to achieve the Paris temperature targets?. *Geophys. Res. Lett.* 43, 7133–7142 (2016).
- 48. Azar, C., Johansson, D. J. A. & Mattsson, N. Meeting global temperature targets—the role of bioenergy with carbon capture and storage. *Environ. Res. Lett.* **8**, 034004 (2013).
- Su, X. et al. Emission pathways to achieve 2.0 °C and 1.5 °C climate targets. Earths Future 5, 592–604 (2017).
- Walsh, B. et al. Pathways for balancing CO₂ emissions and sinks. *Nat. Commun.* 8, 14856 (2017).
- Scott, V., Gilfillan, S., Markusson, N., Chalmers, H. & Haszeldine, R. S. Last chance for carbon capture and storage. *Nat. Clim. Change* 3, 105–111 2013).
- 52. Le Quéré, C. et al. Global Carbon Budget 2015. Earth Syst. Sci. Data 7, 349–396 (2015).

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Author contributions

J.R. coordinated the conception and writing of the paper, performed the scenario analysis and created the figures; J.R., K.V.C., A.P., G.L., J.Em., S.F., E.K., K.R. and D.P.v.V. designed the scenarios, which were developed and contributed by all modelling teams, with notable contributions from S.F., T.H. (AIM/CGE), K.V.C., J.Ed. (GCAM), D.G., E.S., J.D., M.H., D.P.v.V. (IMAGE), O.F., P.H., V.K., J.R., K.R. (MESSAGE-GLOBIOM), J.S., F.H., A.P., G.L., E.K. (REMIND-MAgPIE) and J.Em., G.M., L.D. and M.T. (WITCH-GLOBIOM); all authors provided feedback and contributed to writing the paper.

Competing interests

The authors declare no competing interests.

Additional information

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Methods

Methodological context. The IPCC AR5 assessed pathways that limited radiative forcing in 2100 to 2.6 W m⁻², allowing a higher level during the century³². This level was deemed likely (>66% probability) to limit global-mean temperature rise to below 2°C relative to pre-industrial levels by 210053. There are various motivations to explore even more stringent scenarios. For example, in several regions and particular subsystems, such as tropical coral reefs, the impacts for a global average temperature rise of 2 °C can already be considerably large54,55. Recent research has also reported discernible differences in climate impacts between a world that is 1.5 °C or 2 °C warmer⁵⁶, and these future impacts depend on the evolution of both the climate and the socio-economic system. Our scenarios provide a quantification of these dimensions for worlds that are 1.5 °C warmer, and can serve as a starting point for further research by other communities, such as, for example, the adaptation, water or sustainable development communities. The scenarios presented here are an extension of efforts to provide scenarios for the integrated assessment of climate-change-related challenges5,6,57: the SSP scenario matrix framework7. Studies already use these narratives to explore the actions required to limit radiative forcing in 2100 to levels varying from $8.5 \,\mathrm{W}\,\mathrm{m}^{-2}$ down to $2.6 \,\mathrm{W}\,\mathrm{m}^{-2}$ (refs 5,9-13,16,31,35,47), and their detailed emissions and land-use developments⁵ serve as inputs for CMIP6 ScenarioMIP17, as well as other MIPs19,20.

Modelling protocol. Participating modelling teams were asked to provide scenarios that comply with specific modelling characteristics and that are derived with the same models, model versions and assumptions as used for the SSPs (see below). The modelling protocol consisted of a set of simulations in which total anthropogenic radiative forcing in 2100 is limited to 1.9 W m $^{-2}.$ The limit of 1.9 W m⁻² is evaluated with the simple carbon-cycle and climate model MAGICC (Model for the Assessment of Greenhouse Gas Induced Climate Change)58 in a setup comparable to the initial setup used for the RCPs⁴. The 1.9 W m⁻² limit was selected to result in at least 0.3 °C of global mean temperature increase difference with corresponding 2.6 W m-2 scenarios, which would be consistent with at least 50% of the global land surface experiencing statistically significant changes in temperatures⁵⁹. The 1.9 W m⁻² limit is achieved in the IAMs by adjusting the CO₂-equivalent carbon price. This means that the radiative forcing target is achieved through reductions in GHG emissions and related co-emissions, but not through intentional increases in aerosol emissions or solar radiation management. Scenarios are run for all SSPs available in each respective modelling framework, and with their corresponding Shared Climate Policy Assumptions²¹, which influence the regional and sectorial application of CO2-equivalent carbon prices (see appendices in ref.⁵). Scenarios are labelled with the forcing target identifier '1.9' in combination with the respective SSP identifier, for example, SSP1-1.9 for a 1.9 W m⁻² scenario with SSP1 assumptions. For each SSP, a marker implementation was selected, which represents the characteristics of that SSP particularly well⁵. If appropriate, insights are drawn from a comparison of marker scenarios only. As was the case with RCP and SSP construction, no account of climate feedbacks to human activities and associated emissions is taken into account in the scenarios reported here.

Model participation. Six modelling frameworks were included in this study: AIM/CGE¹¹, GCAM4¹², IMAGE⁹, MESSAGE-GLOBIOM¹⁰, REMIND-MAgPIE¹³ and WITCH-GLOBIOM¹⁶. To ensure consistency and comparability, the study was carried out with the same model versions and setup as used for the other SSP-RCP work⁵. Detailed descriptions of the SSP implementations in all participating frameworks are available as part of a special issue on the quantification of the SSPs^{9-13,16}, with overview papers showing a comparison of results⁵ as well as a synthesis of key insights related to the energy system³⁵ and land use³¹. An overview of model documentation, including the native regional resolution of the models and extensive references, is available in appendix D of ref. ⁵. Supplementary Table 1 provides a succinct overview of the modelling frameworks and key references.

Two modelling frameworks have slight updates to their model setups since earlier publication of the SSP-RCP work⁵: (i) GCAM: The implementation of near-term policy restrictions as dictated by the Shared Policy Assumptions^{5,21} has been modified for 'F2' (see ref. ⁵) by ensuring that a linear carbon–price trajectory is followed between 2020 and 2040. GCAM's agricultural assumptions in 2020 have been adjusted to better align emissions with observations. In particular, agricultural productivity estimates from 2011 to 2020 have been reduced. (ii) WITCH: A recalibration in the supply cost curves of Storage and Transportation of CO₂ has been carried out. On the basis of the regional storage cost curves of ref.⁶⁰, availability curves per region have been fitted to provide better cost estimates as the amount of stored CO₂ increases markedly, and to ensure the estimated storage potential is in line with more recent publications.

Not all modelling teams attempted to model all SSPs, and many only implemented a subset, either because their model was not appropriate to represent the particularities of a specific SSP or because of time and resource constraints. No SSP3–1.9 scenarios have been reported as reaching a 2.6 W m⁻² target was, under these assumptions, already not possible⁵ (Supplementary Table 1, Supplementary Fig. 1, Supplementary Text 2). Marker implementations are available for the $1.9\,W\,m^{-2}$ scenarios for SSP1, SSP2 and SSP5.

ARTICLES

The set of modelling frameworks participating in this study represents an ensemble of opportunity. However, it nevertheless represents a wide variety of modelling approaches and model behaviour. Several different model types are represented, including Computable General Equilibrium (CGE) models, partial equilibrium models and hybrid models that combine a systems dynamics or a systems engineering model with a CGE (see Supplementary Table 1). Three frameworks are intertemporal optimization frameworks, and the other three are recursive dynamic frameworks (see table 1 of ref. ⁵). The set of modelling frameworks spans the whole spectrum of model response classes as identified in ref. ⁶¹, that is, from low (WITCH) to high response (for example, REMIND, GCAM and MESSAGE). Considering these various dimensions, the ensemble of opportunities described by modelling frameworks participating in this study spans a wide diversity of the models that are available.

The scenarios presented here do not consider all potential CDR options (for example, they do not include direct air capture, enhanced weathering, biochar, soil organic carbon or ocean fertilization) and exclude solar radiation management. In these scenarios, CDR is thus mainly achieved by BECCS or afforestation.

Emission and temperature assessment. GHG emissions here always refer to the gases of the Kyoto basket (that is, CO2, CH4, N2O, HFCs, PFC and SF6 but excluding the recently added gas NF3)62, aggregated with 100-year Global Warming Potentials from the IPCC Fourth Assessment Report63. Globalmean temperature change is reported relative to the 1850-1900 base period, here referred to as the pre-industrial period. Exceedance probabilities are computed with a probabilistic setup of the MAGICC model^{64,65} similar to the setup used in the IPCC AR5 Working Group III contribution³². The distribution of equilibrium climate sensitivity assumed in this setup is derived from the climate sensitivity assessment of the IPCC Fourth Assessment Report and hence fully consistent with this Report⁶⁵. Our setup has similar results when updated to the values of the IPCC's most recent assessment (see ref. 66). The implied transient climate-response distribution has a median of 1.7 °C with a 5-95% range of 1.2-2.4 °C. The performance of this model setup is compared to the response of complex general circulation models in fig. 6.12 of ref. 32.

Data availability. Scenario data for all SSPx-1.9 scenarios will be made accessible online via the SSP Database portal: https://tntcat.iiasa.ac.at/SspDb/.

References

- IPCC Climate Change 2014: Impacts, Adaptation, and Vulnerability (eds Field, C. B. et al.) 1–32 (IPCC, Cambridge Univ. Press, 2014).
- 54. Frieler, K. et al. Limiting global warming to 2 °C is unlikely to save most coral reefs. *Nat. Clim. Change* **3**, 165–170 (2013).
- Schleussner, C. F. et al. Differential climate impacts for policy-relevant limits to global warming: the case of 1.5 °C and 2 °C. *Earth Syst. Dynam.* 7, 327–351 (2016).
- 56. Moss, R. H. et al. The next generation of scenarios for climate change research and assessment. *Nature* **463**, 747–756 (2010).
- Meinshausen, M., Raper, S. C. B. & Wigley, T. M. L. Emulating coupled atmosphere-ocean and carbon cycle models with a simpler model, MAGICC6 — part 1: model description and calibration. *Atmos. Chem. Phys.* 11, 1417–1456 (2011).
- Claudia, T., Brian, O. N. & Jean-François, L. Sensitivity of regional climate to global temperature and forcing. *Environ. Res. Lett.* 10, 074001 (2015).
- Hendriks C., Graus W. & Van Bergen F. Global Carbon Dioxide Storage Potential and Costs Report No. EEP-02001 (Ecofys, 2004).
- 60. Kriegler, E. et al. Diagnostic indicators for integrated assessment models of climate policy. *Technol. Forecast. Social. Change* **90**, 45–61 (2015).
- Decision 24/CP.19. Revision of the UNFCCC Reporting Guidelines on Annual Inventories for Parties included in Annex I to the Convention 1–54 (UNFCCC, 2013).
- 62. IPCC Climate Change 2007: The Physical Science Basis (eds Solomon, S. et al.) (Cambridge Univ. Press, 2007).
- Meinshausen, M. et al. Greenhouse-gas emission targets for limiting global warming to 2°C. Nature 458, 1158–1162 (2009).
- Rogelj, J., Meinshausen, M. & Knutti, R. Global warming under old and new scenarios using IPCC climate sensitivity range estimates. *Nat. Clim. Change* 2, 248–253 (2012).
- Rogelj, J., Meinshausen, M., Sedláček, J. & Knutti, R. Implications of potentially lower climate sensitivity on climate projections and policy. *Environ. Res. Lett.* 9, 031003 (2014).
- IPCC Climate Change 2014: Mitigation of Climate Change (eds Edenhofer, O. et al.) 1–33 (Cambridge Univ. Press, 2014).