Societal breakdown as an emergent property of large-scale behavioural models of land use change

Calum Brown, Bumsuk Seo, and Mark Rounsevell

Responses to reviewers (our responses in red)

5 We're grateful to both reviewers for their detailed and insightful reviews, which raise a number of important points and provide helpful suggestions. We have revised the manuscript as described below.

Reviewer 1 (Patrick Meyfroidt)

This manuscript presents a modelling experiment using a large-scale behavioral model of land-use change over Europe. Overall, the research done is very valuable. In itself, it is not a breakthrough, as it builds on many earlier modelling efforts and explores one additional aspect of what can be done with this modelling framework, but it reveals a series of interesting insights.

Thanks for the positive comments.

However, in its present form I don't think that the manuscript is ready for publication. It misses details on some important methodological aspects as well as on some of the results that are at the core of the added value of the paper. These two issues

15 are interlinked, in the sense that without these methodological details it is not really possible to appreciate the value of some of the results, and these results themselves justify some of the methodological progresses.

So my main comment is to clarify key methodological aspects related to the ecosystem services demand and supply calculations, as well as the land use decision-making process, and to better present the results in particular related to the behavioral aspects. Most of my substantial comments are related to this. I return to this more in details below, and add two

20 other substantial comments, on the evaluation of the scenarios based on "shortfall", and on the differences between standard economic models and this effort, as well as some minor ones.

Many thanks for these comments. We fully accept the criticisms and are eager to ensure that the model is interpretable to readers, and have adopted all of the suggestions in the revised manuscript.

1/ On gaps in methodological description and results presentation:

25 In general: I understand that the two modelling frameworks (Crafty and IAP) have been described elsewhere, but in order to understand the added value here more explanations are needed. The behavioral part of the scenarios is very lightly, and unclearly, described. There's 3 lines on p.4 (1.24-26) to introduce the fact that the Agent Functional Types (AFT) have different behaviors, and then 10 lines p.7 (1.11-20) which describe very

vaguely these different behavioral parameters. Beyond this, the basic behavioral and decision-making framework of the agents is not clearly described. The sentences on p.5, 1.24-31 are very unclear to me. The description in the Appendix, in particular

30 is not clearly described. The sentences on p.5, 1.24-31 are very unclear to me. The description in the Appendix, in particular p.31, 1.6-10 (which re-explains, but much more clearly, the p.5 1.29-31), and the caption of Table A4 p.38, are much more clear to me. This should be in the main text. Basically, we need to understand how agents make decision to either maintain or change land use, or how they are outcompeted by others. Just as two examples, in the present version it is impossible to understand

how an agent can be "outcompeted by other agents" (p.7 l.16) or why "the more extreme behaviours [are] being selected out by a competitive process" (p.13, l.29).

We have now substantially extended (and tried to clarify) our presentation of the competition process and behavioural aspects of the model. We have included a new description of how these interact in the methods section (including by moving text from

- 5 the Appendix into the main text as suggested), expanded the analysis of the different behavioural parameterisations in the results section, and given further interpretation of their effects in the discussion section (including the selection of behaviours that we previously suspected was occurring, and which we now show is a genuine but relatively weak, scenario-specific effect). We agree that understanding the operation of the model and the role of agent behaviours within it is necessary for interpreting the results, and hope our explanations are now sufficient.
- 10 This is similar when it comes to the results of this exploration. The results are introduced on p.9, 1.25, but this is in fact referring to the Supplementary Material, and to Table 2 which is just a narrative summary of the results of the different scenarios, including the sensitivity of the behavioral parameters. But no actual result is presented in the main text. To me, this is insufficient.

As this is part of the title of the paper and one key aspect in the paper is to argue that such behavioral models are important to

- 15 explore potential decision-makings that differ from monetary optimization, this would deserve more details. I do understand that one key conclusion is that is is not the behavioral parameters in themselves that matter so much, but rather their basic existence in the model, so that outcomes do not differ so much depending on the behavioral parameters but do differ between this model and others based on neoclassic economics. But still, if you want the reader to buy that idea, you really need to explain much more clearly how do the agents in this model, behaviorally-speaking, differ from basic monetary-optimizing
- 20 agents that are implicit in many other LU models. And you should find a way to present, in the main paper, some of the results of the behavioral exploration. Currently, this is a set of graphs in Appendix 2 which will be, in essence, totally inexistent for most readers. I understand that this si lot of graphs to summarize in perhaps one or two Figures in the main text, and it probably requires some creativity, but I really think that discussing these results without presenting any of them in the main article is not correct.

25 The discussion on this (e.g., p.11 1.5-7) is itself very thin, and sometimes not very clear (e.g. p.13, 1.5-11)...

We follow the general and specific suggestions here by:

- Explaining more clearly the processes by which land use changes in the model (as for the above comment, but also specifically with respect to differences between this approach and purely economic or optimisation-based approaches) – methods section;
- Further analysing the behavioural experiments and including new graphical and textual results of these in the main text and Appendix – results section, especially to show the effects of behavioural variations in the amended plot of supply-demand curves (Fig. 2);
 - 3) Providing more interpretation of these effects in the discussion section;
 - 4) Extending and clarifying the specific sections identified as being thin or unclear.

As noted in the comment, the behavioural variations are purely exploratory at this stage, and we do not wish to place undue weight on them or to overshadow our findings about the importance of 'supra-economic' behaviour in general, but hope that the revisions strike a better balance than did the original text.

5 2/ In addition, I have two other substantial comments:

2.1/ On the evaluation of the scenarios based on "shortfall": One key outcome on which the scenarios are discussed is the relative shortfall between demand and supply. This notion brings ambiguities. As discussed by the authors themselves (p.12), if supply really crashes because of socio-economic collapse, then at some point the demand will fall too, in a way that is not captured in the model. OK, but my concern also goes in the other direction: The scenarios that correspond to high socio-

- 10 economic development most likely generate a higher demand. If there is a shortfall in these higher demands, does that really mean that society's well-being is harmed? Is it possible to consider that some shortfall in a high-demand society reflects something more like a reasonable supply, which may lead to sufficient consumption? To formulate this in a less normative and more technical way, is it appropriate to only evaluate the outcomes in terms of the shortfall between demand and supply, or would it be reasonable to also evaluate the outcomes in terms of the overall (absolute value) of the supply?
- 15 This is a very interesting point, thanks. We now interpret shortfalls and surpluses of food in terms of (approximate) per capita consumption to account for the different demand levels between scenarios, and also discuss scenario implications in light of this point. We also feel that this highlights a general shortcoming in scenario modelling (including our own) in that the distribution of financial capital and hence access to food is rarely considered, and we include this point in the discussion.
- 2.2/ On the differences between standard economic models and this effort: The discussion, p.12 1.9-12, suggests that these economic models would be unable to represent such a collapse. I'm not totally clear about all the reasoning here. These models would indeed (1.10-12) display rising food prices, and thereby some maintenance of food production, but I'm not totally clear on how this would be so different than the results presented here noting that the demand isn't adjusted here, as acknowledged by the authors. p.8, they say: "Conversely, where these capitals declined substantially, widespread extensification and abandonment of land occurred...": Yes, that makes sense, but you would expect this to also occur in standard economic models.
- 25 What is precisely the argument?: That this model predicts a much stronger decline in food production than standard economic models, then this has to be substantiated by numbers, Or that this decline is more realistic than the lower decline in standard economic models, then this has to be justified convincingly. The idea in standard economic models that with production shortfalls, prices would rise, which would thus somehow buffer the production shortfall by mobilizing more capital towards agriculture is reasonable, especially considering that at some point anyone would have to admit that food is a basic need.
- 30 We agree that the reasoning here wasn't clear, and now discuss these issues in substantially more detail. The differences we now highlight between this model and the 'standard' economic paradigm are that:
 - 1) A wider range of ecosystem services are included here, and the value of production of these services is more balanced;

- 2) These services, including food production, are sensitive to a wider range of capitals (including e.g. social and human capitals) this is a major cause of the shortfalls we find in some scenarios, as economic inputs cannot entirely make up for the decline of other capitals;
- 3) In purely economic terms, both satisfactory prices and adequate financial capital are necessary to support modelled production (as calibrated in the AFT production functions), and financial capital here reflects the scenario conditions. Therefore, even if prices were entirely (and, we suggest, unrealistically) unconstrained, scenarios with low financial capital would be vulnerable to shortfalls in production;

4) Food prices here do rise as supply falls behind demand, but they do so according to a specified function that is not designed to guarantee the financial input necessary to achieve equilibrium – i.e. the prices are not artificially set at a level at which modelled shortfalls can be specified or eliminated;

5) The response of the model to price signals is ultimately determined by agent decision-making, which is not forced to be economically optimal – in many cases, small differences in the value of production do not stimulate changes in modelled land use;

6) Because no overall optimisation of land uses is used, the model is path-dependent and can move into states from which more productive or profitable alternatives are hard to reach (e.g. if a large number of marginal agents satisfy demand and/or are unwilling to change their land use, more efficient intensive agents may struggle to come in).

As the reviewer suggests, these differences are not necessarily fixed; many 'standard' economic models could certainly take a similar approach to most of these points. In the revised manuscript we therefore discuss these largely as differences in practice, rather than principle, and emphasise that the findings show the need for clear and well-supported assumptions regarding

20 economic drivers and responses, rather than prioritising a particular approach *per se*.

3/ A few more minor comments:

* Abstract: "economic irrationality": This is an ambiguous formulation. If one sticks to monetary profit optimization (I agree that this can be called "standard" economic models, but this needs to be explicit), lots of behaviors are irrational, if an "enlightened" economist expands a utility function to encompass pretty much anything, then it is hard to find any irrational

- 25 behavior, and so on. (without entering into the whole discussion, things like imitation, sticking to one's behavior, and so on, can be perfectly rational under a given set of information and agentic capabilities). Better rephrase without such a connotation, or perhaps at least talk about irrationality in regards to monetary profit maximization. Next sentence, "this theoretical optimum" bears the same unclear connotations to me. This notion of "irrational" agent comes back later on and is misleading to me. Point taken, and we now avoid the term 'irrationality'.
- 30 * p.2: "... where they are most required; when socio-ecological processes break down...": Yes, but this is only one example, any other situation of regime shift / systemic change / land-use transition brings similar challenges for basic land system models, be they based on economic rules or on statistical calibration.

This is true, and now acknowledged in the text.

* p.8: "..., which were not substantially reduced...": who is this "which"? The following sentence seems to suggest that you
refer to the divergences in land system outcomes, but the sentence is odd as it is not correct to write that "divergent land systems (...) were not substantially reduced".

Corrected, thanks.

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* p.9: 1.1-15: This is described qualitatively. It would be good to find a way to present quantitatively the differences between these scenarios, in a way that would convince the reader of some of the points made, for example that SSP3 has such an impact compared to the climate scenario.

We now include some quantitative comparisons between the scenarios as suggested.

5 * p.11: 1.14-17: Maybe yes, maybe no. This depends on how is the actual balance between ES, compared to your own way to balance them. But still I agree with the conclusion of the following lines that better understanding how these trade-offs are actually formulated in reality is crucial.

We have rephrased to emphasise the latter point.

- * p.11: 1.23-24: and this would likely reinforce the shortfalls, right?
- 10 Correct, added.

Reviewer 2 (Gunnar Dressler)

In this manuscript, the authors present an intriguing example of an agent-based model applied at the European scale that is able to produce realistic long-term projections of land-use change. The model - CRAFTY-EU - is an extension of the

- 15 CRAFTY model that has been developed by the same author group. The main aim of the paper is therefore not to comprehensively present the model itself, but to rather to deliver a "proof-of-concept" that such an agent-based model a) can be applied at a large spatial scale and b) although it surely simplifies several relationships, e.g. regarding the representation of land-users, is well suited to explore long-term land-use change dynamics. It actually allows more flexibility in this regard, as it does not impose any constraints regarding optimality of the emerging land-use patterns and is therefore also able to simulate
- 20 a although undesirable breakdown of the land system.

Overall, the manuscript is well-written with regards to wording and grammar, and as such is pleasant to read. Especially the Introduction and Discussion Sections are clearly structured and the argumentation, backed by a range of references, is sound. However, the description of the model itself, and parts of the results, require some improvement, as they are not always fully and clearly described - which is also my main critique of the manuscript.

25 Thanks for the positive comments and identifying these shortcomings (which we note largely agree with those identified by Patrick Meyfroidt).

Parts of the manuscript – especially in the Methods & Results Sections – are a bit hard to follow, as the manuscript makes ample references to the Appendix here (11 references to the Appendix in Section 2, 6 references to the Appendix, respectively results in the Appendix in Section 3). At some points, it is therefore hard to follow the manuscript without jumping back and

30 forth, respectively without fully reading the Appendix, in addition to the main text.

Yes, we accept this point. We now include more detail from the Appendix directly in the main text (methods section especially), and have also expanded the description of how the model operates as outlined above. These revisions are intended to minimise the need for readers to refer to the Appendix in order to understand the results we present. Examples in the Methods section are: • p. 4, l. 24-26: "Behavioural differences between AFTs (in terms of willingness to change land use or abandon land, and range of variations in capital sensitivities and ecosystem service production levels) were also introduced to assess the robustness of model outcomes to behavioural variations (see below and Appendix A)" how these behavioural variations between AFTs is implemented is not clearly evident from the manuscript, there's only a very brief explanation on how behavioral parameters

5 are varied at the end of Section 2.6 (p. 7, l. 11-14). As the authors stress that the representation of behavioral differences between AFTs is a crucial aspect of the model, some more information on this should be added to the main text. We have now substantially revised and extended this section (we particularly focused on this as both reviewers made similar points here).

• p. 5, l. 23-24: "with the agents producing the most (or the most highly valued) services gaining the highest benefit values and

10 therefore best-placed to win the competition for cells (Appendix A)" from this, it is not clear how agents compete for cells – in the appendix.

This has also now been expanded and clarified.

In the Results Sections, this applies particularly to the model evaluation in Section 3.1, where the simulations starting with no initial land use map and with the baseline map are described. If these results are considered important, then the corresponding

- 15 figures should also be included in the main text otherwise, they should only be addressed in the Appendix. We are not entirely convinced of this – we believe it's important for readers to know that the model behaved in this way (achieving a reasonable outcome when initiated without a starting map, and remaining static when run under static conditions) because it speaks directly to the interpretation and reliability of the main results. However, we're not sure that the figures themselves merit inclusion in the main text, as they don't really reveal anything further. We therefore keep the existing structure
- 20 for now, with the figures available in the Appendix for readers who wish to check the claims we make in the text about evaluation.

Questions regarding the Results:

• Is there any explanation for the fact that "the model spontaneously produced realistic land use configurations on the basis of land productivities, AFT parameterisations and demand levels" (p. 8, l. 8-10) – i.e. the strength of external forcings?

- 25 We now explain this result in slightly more detail in the text, to say that we would expect some level of agreement with reality because the model was given capital levels and demand levels, which are the major drivers of land use. However, we would not expect complete agreement because there are numerous potential configurations of land use that would represent 'reasonable' outcomes, and a model unconstrained by an initial map should be able to achieve many of these. While it is hard to be precise in the interpretation of this result, we suggest that it shows that the use of capitals and demands allows at least
- 30 the possibility of reproducing observed land use dynamics i.e. that the model responds to these basic drivers broadly as real land systems do.

• What are the main reasons for the "widespread extensification and abandonment of land occurred and large shortfalls in service levels" (p. 8, l. 24) – is it because agents only decide about profit?

This is also similar to comments above about the role of economic and non-economic behaviour in the model. As outlined there, we have substantially extended our discussion of these findings to explain these outcomes – which develop largely because agents are reliant on a range of capitals (natural productivity, human, social, manufactured and financial capitals), several of which decrease dramatically in some scenarios. This leaves agents unable to achieve a level of service production

- 5 that brings returns (economic & non-economic) defined as acceptable.
 - I did not fully understand the "peak effect" (p. 10, l. 1) with regard to individual behavioral variations and how irrational agents are selected out can you explain this effect in more detail?

This point was largely speculative in the initial manuscript, but in response to this suggestion we have looked into the evolution of behaviour in more detail. We found that behavioural parameters do systematically change in some scenarios, and describe

10 and attempt to explain this in the text and in new figures in the Appendix. This (slight) effect seems to be due to the persistence of agents with tolerance of low benefit values and unwillingness to change land use in more challenging scenarios (particularly SSP3).

• Figure 1: Where do the sharp transitions in the mean capital value plots at specific points in time come from?

These come from the timesteps at which capital values were available from the IAP (between which values were linearly

- 15 interpolated). We have added and explanation to the figure legend.
 - In the Discussion Section, the authors address both aspects regarding the model design (and its limitations), as well as regarding the model results. Although the authors clearly state that the current model represents a "a first-step towards improved understanding of behavioural processes within large-scale land systems." (p. 10, 1. 30), some of the simplifications that the model assumes could be addressed in more detail. One of the main limitations of the model is its relatively coarse resolution
- 20 of only 10 arcminutes, and with that the limitation to only being able to represent generic agent functional types, rather than individual land managers. As such, the model is not able to represent individual land user's decisions or interactions. This does not diminish the results of the current study, but I would expect some statements here on a) how the focus on generic AFTs only vs. individual land managers may influence results and b) how a more detailed representation of individual land users may be achieved in future versions of the model.
- 25 Thanks for the suggestion, and we agree that these are important considerations. We now discuss both of these points in the discussion section, in light of this application as well as basic CRAFTY design (based on the premise that an informative balance between important behaviour and 'noise' exists somewhere above the individual level e.g. at the level of 'functional types'), as well as other models' findings and discussion in the literature.

Questions regarding the Conclusions:

30 • You state that "behavioural effects may be partially 'self-correcting', with more extreme behaviours being selected out by a competitive process." (p. 13, l. 28- 29) – could show this, e.g. by plotting the distribution of behavioral parameters at the beginning and end of the simulation?

Thanks for this suggestion. We have done this, and include the resulting plots in the Appendix with further discussion of the effect in the main text.

Technical corrections:

• p. 2, l. 30: "If a new generation of behavioural models are to make. . . " : are is

Corrected, thanks

• p. 13, l. 18: "CRAFY-EU"

5 Corrected, thanks

• Figure 1: please label the subfigures, e.g. a, b, c, d

Done

• Figure 2: I found it very unintuitive to read the plot with negative values indicating a surplus, and positive values indicating a shortfall – could you show supply / demand instead, which would make it much easier to read over-/undersupply? Also, this

10 would align better with Figure B2 in the appendix.

Changed as suggested (also with the addition of ranges to show the effects of behavioural variations).

• Overall: in places, where references are cited with "e.g.", these references are enclosed in double brackets, e.g. p. 2, l. 23-35 Thanks, now corrected

Appendix A & B: As the appendix includes a number of information that are quite essential to understand the manuscript, I

15 would also suggest a number of minor corrections here:

1. Ordering of tables: You refer to Table A2 in the text only after referring to Tables A3 & A4 (p. 30, l. 19) – please correct the numbering.

Thanks; this table should have been referred to earlier, and now is.

2. Table A3: You refer to standard deviation values formatted in red, but table values are only black, this should be corrected.

20 Also, the maximum service production levels are quite different from each other – what's the unit/scale of each service?

We have corrected the tables to refer to the bracketed rather than red values, and to include the standardised production values that form the basis of the model calculations. The original production levels do have very different units and are a mix of empirically-based values (e.g. extractive goods) and abstract values (e.g. recreation).

3. Table A4 that contains all the details on the variation of the behavioral parameters is only explained in the table caption -

25 there's no explanation of the parameter variations in the text at all. This should definitely be added. Also, the Name column shows only abbreviations of the AFT names, which aren't used anywhere else – please use the full names here, as it the table is otherwise difficult to read and the column width is not limiting factor.

We have made these changes, and added the paramerisation explanation to the main text, Section 2.6.

4. p. 31, l. 5-6: use subscript or _ for better readability for ms, us, rs, i.e. m_s, u_s, r_s

30 Thanks for spotting this, we have changed to subscripts.

5. Figure B2 a) + b): readability of the figure would be improved if you add a grid to the figure, so that numbers of agents could be more easily compared, b) add a 1.0 line to make it easier to see whether an over-/undersupply occurs.

Done, thanks for the suggestions

6. Figure B4: at first, I was wondering whether this figure made sense here, as numbers of agents per AFT did stay constant over the whole simulation run (as per definition of the behavioral parameters in Table A4, where giving up/giving in thresholds are 0 in the baseline run) - only after careful examination I noticed some tiny changes towards the end of the simulation, e.g. for the Intensive Pastoral AFT - could you highlight these changes in the graph? Otherwise, I would omit this figure from the measurement as it does not account of the substantial here fit.

5 manuscript, as it does not provide a substantial benefit.

It's correct that the plot shows effectively no change (agent numbers are almost completely constant, as described in section 3.1.2), but we find this an important if not particularly dramatic result. The giving up and giving in thresholds do not preclude change here – while agents will tolerate any positive benefit (giving up threshold), they will also change to a land use with any advantage (giving in threshold) – i.e. they are actually highly sensitive to competitive advantages. As a result the plot

10 demonstrates model stability under static conditions rather than a predetermined result of parameterisation, and we believe showing this visually in the appendix may be a useful addition to saying so in the main text.

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Societal breakdown as an emergent property of large-scale behavioural models of land use change

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Abstract. Human land use has placed enormous pressure on natural resources and ecosystems worldwide, and may even prompt socio-ecological collapses under some circumstances. Efforts to avoid such collapses are hampered by a lack of knowledge about when they may occur and how they may be prevented. Computational models that illuminate potential future developments in the land system are invaluable tools in this context. While such models are widely used to project biophysical changes, they are currently less able to explore the social dynamics that will be key aspects of future global change. As a result, strategies for navigating a hazardous future may suffer from 'blind spots' at which individual, social and political behaviours divert the land system away from predicted pathways.

- 15 We apply *CRAFTY-EU*, an agent-based model of the European land system, in order to investigate the effects of humanbehavioural aspects of land management at the continental-scale. We explore a range of potential futures using climatic and socio-economic scenarios, and present a coherent set of cross-sectoral projections without imposed equilibria or optimisation. These projections include various behavioural responses to scenarios including non-economic motivations, aversion to change, and heterogeneity in decision-making. We find that social factors and behavioural responses have dramatic impacts on
- 20 simulated dynamics, and can contribute to a breakdown of the land system's essential functions in which shortfalls in food production of up to 56% emerge. These impacts are largely distinct from, and at least as large as, those of projected climatic change. We conclude that the socio-economic aspects of future scenarios require far more detailed and varied treatment. In particular, <u>deviation from simple economic rationality the extent of economic 'irrationality'</u> at individual and aggregate scales may <u>determine-profoundly alter</u> the nature of land system development and the achievability of policy goals, with established
- 25 pathways being highly vulnerable to deviation from this theoretical optimum.

1 Introduction

10

Human use of land resources has led to transformation of much of the Earth's surface (Hooke and Martín-Duque, 2012; Pongratz et al., 2008; Ramankutty et al., 2008). This transformation has enabled rapid rises in human population sizes and some living standards, but has also been a driving force of climate change and mass extinction (Newbold et al., 2016; Steffen

30 et al., 2015). These consequences have become so severe that they threaten the continued provision of many of the essential 'contributions to people' that terrestrial environments make (Díaz et al., 2018). Societies now face the enormous challenge of sustaining these contributions while simultaneously overcoming ingrained inequalities in their distribution (United Nations, 2017).

Computational models play a crucial role in understanding global change and identifying strategies to avoid its worst impacts. However, the systemic complexity that makes these models essential also makes them difficult to verify, inevitably incomplete

- 5 and therefore of limited accuracy (Beven, 2007; Brown et al., 2016a; Smith, 2001). Indeed, recent research suggests that land system models tend to produce unrealistic and inconsistent projections of human behaviour in particular (Alexander et al., 2017; Brown et al., 2019; Searchinger et al., 2017; Turner et al., 2018). This may make these models inapplicable in exactly the circumstances where they are most required; when socio-ecological dynamics cause systemic change, regime shifts or processes-break-downs and systems collapse (Cumming and Peterson, 2017).
- 10 One necessary improvement in modelling practice is the adoption of a wider range of conceptual and technical approaches (Alexander et al., 2017; Huber et al., 2018; Meyfroidt et al., 2018). At present, a small number of simplifying assumptions have become standard in land system modelling, allowing models to operate over large geographical extents and thematic areas without becoming computationally intractable. Broad assumptions about human behaviour are particularly common, usually following a paradigmatic reductionist approach that emphasises the role of macro-economic drivers of land use change
- 15 (Brown et al., 2016a, 2017; Calvin and Bond-Lamberty, 2018). These assumptions tightly constrain the representation of human decision-making, often forcing requiring it to adhere to exogenously imposed equilibria. Furthermore, a focus on the agricultural sector has meant that other sectors (e.g. forestry, urban development) have generally been treated as separate systems rather than interacting components of the land system as a whole (Brown et al., 2017; Smith et al., 2010; van Vliet et al., 2019).
- 20 These shortcomings particularly constrain exploration of the effects of the social aspects of future scenarios, which, while often quite dramatic, are not reproducible through the predominantly biophysical parameters of most land use models (Müller-Hansen et al., 2017; Riahi et al., 2017). Alternative, well-supported conceptualisations of the human land use system are available, and some have been formalised in agent-based or behavioural models that focus on individual-level decisions from which system properties emerge (e.g. (Arneth et al., 2014; Brown et al., 2016b, 2017; Fagiolo and Roventini, 2017; Magliocca,
- 25 2015; Rounsevell et al., 2012b). To date, these models have been limited in scope, mainly operating only in specific contexts or over small geographical areas (e.g. (An, 2012; Brown et al., 2017; Robinson et al., 2018)). However, their focus on underlying processes makes them suitable for scaling out and scaling up, across entire, coherent land systems (Rounsevell et al., 2012a). Recent conceptual and technical developments make this scaling feasible (Arneth et al., 2014; Verburg et al., 2015), and associated studies suggest that micro-scale behavioural processes can have significant macro-scale effects (Bai et al.)
- 30 al., 2016; Blanco et al., 2017a; Brown et al., 2018b; Calvin and Bond-Lamberty, 2018). If a new generation of behavioural models are-is to make a substantial contribution to Earth System modelling, they-it must satisfy a number of requirements. First and foremost, they-models must achieve accuracy in their representation of basic processes that transcend land sectors, geographical areas and scenario conditions. Given this, models can move beyond context-specific pattern-matchingcalibrations and retain sufficient flexibility to explore land system development under

uncertain future global change. By the same token, these models need to incorporate relevant decision-making processes at a range of scales, from individual to community and government, so minimising the role of exogenous and potentially inconsistent assumptions about nested actions (Galaz et al., 2012; Lippe et al., 2019; Rounsevell et al., 2014). Beyond behaviour, models must also reflect the true range of land use options, including gradients from subsistence production to

5 profit-maximisation, highly extensive to highly intensive management, and entirely uni-functional (monocultural) to highly multifunctional or mosaic land systems (McDermid et al., 2017; Verburg et al., 2012). In order to move towards these goals, we have developed CRAFTY-EU, a continental-scale, agent-based model of the European land system based on the CRAFTY modelling framework (Murray-Rust et al., 2014). We describe the design,

calibration and evaluation of this model before using it to explore future developments in Europe's land system under a range

10 of climatic and socio-economic scenarios. We assess the sensitivities of these developments to scenario conditions and various forms of land manager behaviours, and their implications for the supply of a range of ecosystem services and land system stability. We then discuss the possible impacts of human behaviour within the land system, as well as the value of novel modelling approaches of this kind for understanding and managing Earth System change.

2 Methods

- 15 CRAFTY-EU is an application of the CRAFTY framework for agent-based modelling of land use change (Murray-Rust et al., 2014; Blanco et al., 2017a; Brown et al., 2018b; Holzhauer et al., 2019; Murray Rust et al., 2014). The CRAFTY framework allows land use outcomes to be modelled as the result of decision-making and competition among individual agents, each of which can represent an individual or multiple land managers, and which produce a range of ecosystem services. Production levels are determined by the productivity of the land (defined through a range of natural and anthropogenic capitals, as
- 20 described below), the intensity of land management, and agents' willingness or ability to produce certain ecosystem services. Agents are grouped into Agent Functional Types (AFTs) (Arneth et al., 2014) on the basis of their management intensity and decision-making characteristics, such as degree of focus on profit-generation and desire to maintain an existing land use. Variation within AFTs allows for individual differences in production levels and land management decisions. Therefore, the model allows for emergent land system properties that are not constrained by assumptions about <u>economic</u> optimality,
- 25 equilibrium or economic rationality, in the sense of satisfying generic economic conditions. The main components of the applied model are summarised in dedicated sections below.

CRAFTY-EU is calibrated using outputs from the IMPRESSIONS Integrated Assessment Platform (IAP), a cross-sectoral, multi-model tool for simulating European land system change (Harrison et al., 2015, 2019; Holman et al., 2017). All necessary input data (described below) are derived from this source, ensuring the transparency and internal consistency of the

30 implementation. This model pairing also allows socio-economic and climatic scenarios to be defined on the basis of comprehensive, cross-sectoral simulations of the European land system that have been extensively evaluated, validated and utilised (Brown et al., 2014a; Harrison et al., 2012, 2016, 2019; Kebede et al., 2015; Pedde et al., 2019b). Changes in the

modelled land system are therefore attributable either to CRAFTY model dynamics (investigated below) or scenario conditions, rather than internal inconsistencies in input data from different sources. Full details of the calibration of CRAFTY-EU are given in Appendix A.

2.1 European application

- 5 CRAFTY-EU covers the European Union-27 (EU member states that include the UK, but exclude Croatia) together with Norway and Switzerland. The model operates at a 10' (arcminute) resolution, with 23,871 grid cells in total. This resolution was selected for its consistency with input data, all of which had the same resolution, for its low computational demands, allowing multiple model runs to be carried out quickly, and because of a shortage of appropriate calibration data at finer resolutions. Nevertheless, this resolution is relatively coarse for an agent-based model application, and means that modelled
- 10 agents cannot be seen as representative of individual real-world land managers in most cases. Instead, they are drawn from semi-aggregated AFTs designed to represent coherent localised land use systems (Letourneau et al., 2012), with management and behavioural characteristics expressed at appropriate generality, as described below.

2.2 Agent Functional Types

Agent Functional Types used in CRAFTY-EU were designed to provide generic coverage of the major sectoral and cross-

- 15 sectoral land systems at local (10') scale across Europe. Key distinctions were made between levels of management intensity and between the ranges of ecosystem services produced (Arneth et al., 2014; Letourneau et al., 2012; Murray-Rust et al., 2014; Paul et al., 2017). The final typology was intended to capture the primary form of land management within each grid cell, while allowing for secondary land uses and variation in local land management practices (Table 1; Appendix A). The initial distribution of these AFTs across the modelled land surface was based on the distribution of land use categories modelled by
- 20 the IAP under baseline conditions (Appendix A), ensuring consistency across initial simulation conditions, and comparability with subsequent scenario-based changes.

The abilities of different AFTs to utilise capitals and produce ecosystem services were defined via capital sensitivity and productive ability parameters (summarised in Table 1 with further details and exact parameterisations in Appendix A; see also Murray-Rust et al. 2014). Where possible, values were derived from simulated production data in the IAP, and otherwise

- 25 assumed on the basis of land management intensity and diversity. Behavioural differences between AFTs were also introduced to assess the robustness of model outcomes. These differences, described in Section 2.6, affected agents' ecosystem service production levels and their willingness to change land use or abandon land, and were intended to approximate differences in productive abilities, priorities and sensitivities between and within agent types. (in terms of willingness to change land use or abandon land, and range of variations in capital sensitivities and ecosystem service production levels) were also introduced to
- 30 assess the robustness of model outcomes to behavioural variations (see below and Appendix A). Urban land use was not actively modelled, but constrained to follow the results of the IAP, which includes advanced modelling of urban development (Terama et al., 2019).

2.3 Land productivities (capitals)

The productive potential of each modelled grid cell was described via five capitals: natural capital (crop productivity, grassland productivity, forest productivity), human capital, social capital, manufactured capital and financial capital. Each capital was

- 5 derived from the IAP as described in Appendix A. Scenario-specific changes in capital values were produced by running the IAP under each scenario in turn and repeating the derivation process. Each of the productivity capitals accounts for climate-induced changes during the period of simulation, including effects of changes in temperature, precipitation and CO₂ levels. These changes were simulated for the IAP by combinations of global and regional climate models: EC_Earth/RCA4 for RCP2.6, and HADGEM2-ES/RCA4 for RCPs 4.5 and 8.5 (Harrison et al., 2019). Socio-economic conditions (as defined by
- 10 the SSPs; (Riahi et al., 2017)) affected anthropogenic capitals (human, social, manufactured and financial) as determined via a stakeholder-led elaboration of scenario narratives and a subsequent uncertainty-based quantification (Harrison et al., 2019; Pedde et al., 2019). Because IAP outputs were only available at three timeslices (2020s, 2050s and 2080s), capital values were linearly interpolated to give annual values for each grid cell over the period 2010-2086.

15 2.4 Ecosystem services, demand and supply

20

The CRAFTY framework is designed to account for the demand and supply of a range of ecosystem services, and we incorporate a representative group for which calibration data are available or for which assumptions related to calibrated land management can be made: timber, meat, crops, carbon sequestration, landscape diversity and recreation. Annual demand levels for each of these services were derived from IAP outputs, via conversion of simulated land cover to service production levels as described in Appendix A. All demand levels are available in Supplement 1.

- Demand levels were converted to 'benefit' values for each agent and cell throughout the simulations, takingthat describe the relative benefit gained for production of each service by simulated agentsaccount of current supply-demand gaps, assumed societal valuations of each service, and agents' abilities to provide each service on each cell (Appendix A). In this case, linear functions of supply-demand gaps were used to calculate benefit values, and these functions were calibrated to ensure equal
- 25 relative valuation of services (i.e. the production of an equal proportion of unmet demand was assigned an equal benefit value whatever the service). This created a balanced competition between agents that was not skewed towards any particular service(s), with no benefit accruing from production when there was no unmet demand, prompting production under shortfalls but not under surpluses.

Resulting benefit values were then used as a basis for competition between agents for ownership of cells. - These values were

30 calculated for each agent and each cell, and used as a basis for competition for land between agents, with the aAgents producing the most (or the most highly valued combinations of) services with the highest benefit values were best-placed to win this competition, but did not necessarily do so because existing agents could choose to persist with their existing land use rather than submit to competition. This process was controlled by 'abandonment' and 'competition' thresholds describing the ranges of benefit values within which agents would abandon their land, persist with their existing land use, or cede their land to another agent (representing an alternative land use) gaining the highest benefit values and therefore best placed to win the competition for cells (Appendix A) (Holzhauer et al., 2019; Murray-Rust et al., 2014). These thresholds were initially fixed

- 5 across agent functional types, and then varied at typological and individual levels as described below. -ThereforeHowever, 'rational' competition was not enforced, meaning that agents with the highest benefit values were not necessarily allocated land, depending on decision-making parameters (outlined below). In addition, agents did not optimise their land uses according to benefit values, and these values were not used-were not designed to ensure full supply of each service. Instead, benefit values, but only to responded in defined ways to changes in demand and supply levels, stimulating production, but not
- 10 guaranteeing a given production level. The model therefore contains no assumptions that override the emergence of suboptimal or non-equilibrium outcomes from scenario conditions.

Another key feature was that demand levels were normalised to produce the same benefit for supply of each proportional unit of unmet demand. This means that production was assigned the same value at any given level of unmet demand for each service. Service production in any part of the EU contributed to satisfying demand levels, representing an assumption of free

15 trade across the modelled area (constrained by the infrastructure and transportation networks described in the manufactured capital values). This is a reasonable assumption given that the EU is a free trade zone. A full description of the valuation and competition process is given in Appendix A.

2.5 Model evaluation

- The CRAFTY modelling framework has been extensively evaluated and applied in previous studies (e.g. (Alexander et al., 2017; Blanco et al., 2017a; Brown et al., 2014b, 2018b; Holzhauer et al., 2019; Murray-Rust et al., 2014)), as has the IAP upon which this application of CRAFTY is based (e.g. (Brown et al., 2014a; Harrison et al., 2016; Holman et al., 2017; Kebede et al., 2015)). Both sets of evaluation have included sensitivity and uncertainty analyses (Brown et al., 2014a, 2014b, 2018b; Kebede et al., 2015; Synes et al., 2018), comparisons to empirical data and to the results of other models (Alexander et al., 2017; Blanco et al., 2017a), full descriptions of model design and functioning (Harrison et al., 2015; Murray-Rust et al., 2014)
 and full, free access to the models themselves including interactive online systems for exploring model outputs (Holzhauer et al., 2016; IMPRESSIONS Project, 2018); https://landchange.earth/CRAFTY). Both models have also been extensively used in, and informed by, a stakeholder engagement process that has occurred over several years across the EU (Kok et al., 2018). Here, additional model evaluation focused on the behaviour, stability and interpretability of the European application of CRAFTY. These characteristics were primarily assessed through two sets of runs under static, baseline conditions, starting
- 30 from an unassigned (empty) land use map and from the baseline land use map derived from the IAP. The purpose of these two exercises was, respectively: 1) to check whether baseline conditions would generate a 'realistic' land use configuration purely on the basis of capital levels and AFT characteristics (i.e. in the absence of any spatial information about land management), and 2) to check for divergence in outcomes from a common starting point consistent with the starting point of other scenario

runs. Model dynamics were checked visually and statistically, using the numbers of agents within each AFT and levels of service provision. Both evaluation exercises are described in detail in Appendix B.

2.6 Simulation schedule

CRAFTY-EU runs on annual timesteps at which a proportion of cells are subject to potential abandonment, adoption, or

- 5 competition (Murray-Rust et al., 2014). In the first evaluation exercise, the model was run over 800 timesteps, with 20% of cells being randomly selected for potential change (i.e. the maximum number of cells that could change at each time step, if required by the competition process). This arbitrary but high rate of competition allowed for rapid changes to the simulated land system, ensuring model dynamics could be clearly perceived. The period required for the model to reach a steady-state was identified, and 10 further independent simulations were then run to this point using different random number generator
- 10 seed values. The second evaluation exercise was performed over 100 timesteps, again with a 20% rate of cell selection. This exercise was designed to run a sufficient number of replicates to identify and understand any divergence from stationarity in model dynamics.

Following the evaluation exercises, simulations were run for 71 timesteps, representing the period 2016-2086, with 5% of cells selected for potential change at each of these timesteps. As an upper limit, this rate is up to an order of magnitude greater than

- 15 observed (e.g.-(Loveland et al., 2012)) and projected land use changes (e.g. (Schmitz et al., 2014)), allowing for the majority of potential changes to be rejected while maintaining scope for rapid land use change under extreme scenarios. These simulations all began from the baseline land use map (Fig. A1.2), and proceeded according to scenario conditions in terms of ecosystem service demand levels and capital values (Appendices 4 and 5). Seven distinct scenarios were simulated, each of which comprised a combination of Representative Concentration Pathways (RCPs) and Shared Socioeconomic Pathways
- 20 (SSPs) (O'Neill et al., 2017) as described in Table 2. Socio-economic scenarios were developed from the SSPs through a stakeholder-engagement process described in detail in Kok et al. (2018). These scenarios were first run through the IAP as described above in order to produce representative levels of capitals and demands for use in CRAFTY-EU. Throughout each simulation, land use maps, numbers of agents in each AFT, ecosystem service production levels and fragmentation indices (fractal dimensions) were recorded.
- 25 For each scenario, five distinct parameter sets were applied to assess the effects of variations in agent's modelled behaviours (full parameterisations and explanations for each of these are given in Appendix A). These parameter sets differed in terms of the abilities of agents to produce services and their tolerance of low benefit values and of competition, all of which varied at <u>AFT and individual agent levels</u>. These variations were designed to represent general behavioural effects arising from land managers' decision-making, accounting for aggregation to the model's spatial resolution, but with a focus on sensitivity
- 30 analysis across a large range rather than calibration at levels thought to be realistic. Parameter set 1 represented a behavioural 'baseline' in which agents responded directly to benefit values with no additional individual or typological behaviour (i.e. the land use with the greatest benefit value was adopted in each case). Parameter set 2 ('increased thresholds') introduced differences in abandonment and competition thresholds to induce abandonment of land when benefit values fell below the

abandonment threshold, and persistence of an existing land use unless an alternative had an additional benefit value of at least the competition threshold. Intensive land use agents were parameterised to be less tolerant of low benefit values, and more willing to switch to a land use with higher benefit values. Parameter set 3 ('individual variation') introduced differences between individual agents, in terms of their levels of production of different ecosystem services as well as their abandonment

5 and competition thresholds. In this case, individual values were randomly assigned from a Gaussian distribution centred on the mean value for each type (Appendix A). Parameter sets 4 and 5 replicated sets 2 and 3 respectively, but with higher values for thresholds and variations (up to 10 times higher, with thresholds of the same order of magnitude as mean benefit values).

These parameter sets differed in terms of the abilities of agents to produce services and their tolerance of low benefit values and of competition, all of which varied at AFT and individual agent levels. These variations were designed to represent general behavioural effects arising from land managers' decision-making, accounting for aggregation to the model's spatial resolution. Under the 'baseline' behaviours, agents persisted with land uses that provided benefits unless outcompeted by other agents, and did not vary at individual level. In other parameter sets ('increased thresholds') agents were less tolerant of low benefit values and competition (i.e. required larger returns to continue their land management or to switch to another) and varied individually in their tolerances and service production levels ('individual variation'). In each case, relatively small and large

deviations from the baseline parameter values were used.

3 Results

3.1 Model evaluation

3.1.1 Simulations with no initial land use map

- 20 Simulations initiated under all baseline conditions, but without the initial land use map, were found to quickly converge to an approximate steady-state (Fig. B1), but not to achieve formal stationarity over 800 timesteps (Box-Ljung test p-values <0.01 for numbers of agents belonging to each AFT and service production levels over 50-timestep periods). This appeared to be due to path-dependent oscillations (over short- and long-timespans) that, while statistically significant, were small relative to total agent numbers and rarely affected the relative rank of each AFT (Fig. B1). These oscillations were amplified by the high
- 25 rate of competition for cells allowed in the evaluation simulations (20% of cells at each timestep), and as such remained broadly in line with expectations, with no evidence of either ongoing systematic change or dramatic regime shifts. The 300th timestep was chosen as representative of model outcomes following the initial period of rapid change, and the numbers of agents belonging to each AFT at this point in each of the ten independent simulations were then plotted (Fig. B2a) along with the proportional supply levels of each service (Fig. B2b). These results showed strong convergence between
- 30 simulation outcomes, with both relative and (approximate) absolute numbers of agents being reproduced in each simulation. Service levels remained between 95% and 110% of demand levels in all cases. In these relatively unconstrained circumstances,

the model tended to produce a slight excess of meat and carbon sequestration services, with a predominance of multifunctional AFTs and a relative lack of intensive-management AFTs. However, aggregated AFTs showed not only spatial consistency across the simulations, but also agreement with the (unutilised) baseline map (Figure B3), suggesting that the model spontaneously produced realistic land use configurations on the basis of land productivities, AFT parameterisations and

5 demand levels. These major drivers of land use therefore appear to have similar, decisive effects in the model as in the real world, without preventing the model from producing reasonable counterfactual land use configurations in the absence of an initial map.

3.1.2 Simulations from baseline map

10 The simulation initialised with the baseline land use map and run under static conditions remained stationary (Box-Ljung test p-values >0.1 for numbers of agents belonging to each AFT and service production levels) (Fig. B4). The total number of agents within each AFT barely changed, with the maximum range in number of agents over the course of the simulation being 2. Further realisations were not generated given this lack of variation and the model stability that it demonstrated under static conditions.

15 3.2 Scenario simulations

health impacts of under-production.

Scenario simulations showed widely considerable divergence betweent land systems being produced by in the mid-2080s under different scenario combinations, which and this were was not substantially reduced altered by behavioural variations between agents (Table 2, Supplement 2). These differences were primarily driven by socio-economic scenario conditions, but also by different levels of climate change between the three climate scenarios used (Figs. 1 & 2). Broadly, where socio-20 economic capital levels were maintained or increased, the land system diverged from the baseline scenario by a relatively limited amount, with widespread intensive management of land and small shortfalls or surpluses of most modelled services. Conversely, where these capitals declined substantially, widespread extensification and abandonment of land occurred and large shortfalls in service levels developed (Fig. 1, Table 2, Supplement 2). These dynamics were partly ameliorated by increases in productivity in some areas associated with high-end climate change, particularly especially north-western Europe. 25 Of particular importance were manufactured and financial capitals, which increase greatly (up to 250%) in some scenarios (e.g. SSP1) and decrease (by around 90%) in others (e.g. SSP3), depending on scenario storylines (Fig. 1 & Table 2). These capitals are crucial in supporting intensive land management in CRAFTY-EU (Appendix A), and so determine the scope for the most productive uses of land. Where these capitals increased, surpluses of services (especially food) developed, and where they decreased, shortfalls developed, reaching 56% of food demand in the RCP4.5-SSP3 scenario combination (Fig. 2). These 30 capital dynamics are also linked to reduced per capita food consumption in SSP3 (Table 2), suggesting that shortfalls in this context translate more directly into hunger than they do in other scenarios where over-consumption provides a buffer to any We simulated three socio-economic scenarios in different climate scenarios, and all showed notable similarities between climates. SSP1 had the most consistent clearest differences; this scenario has high demands for all services, and the difference between climate scenarios was due to increases in average crop and forest productivity capitals under RCP4.5 relative to

5 RCP2.6. These productivity changes increased the competitiveness of intensive management enough to allow it to outcompete more extensive, multifunctional land uses, and so allowed production to <u>meet 15-30% more of demand than under RCP2.6</u> increase enough to satisfy demand.

The most consistent and most negative scenario was SSP3, in which economic and social challenges led to disintegration of the land system across much of Europe, with large areas being abandoned, managed extensively, or fluctuating over time

- 10 (Figure 1, Table 2, Supplement 2). These dynamics were particularly pronounced in more fertile areas of Europe, where currently dominant intensive management declined dramatically during the first half of the century. <u>As a result, Similar-similar shortfalls of almost 50% of food demand results</u>-were found <u>between 2050 and 2080</u> in both RCP4.5 and RCP8.5, suggesting that changes in climate were minor in comparison to the almost complete loss of financial and manufactured capitals that undermines the productive use of land in SSP3. Nevertheless, supply levels increased markedly towards the end of the century
- 15 in RCP8.5, as increased natural capitals (i.e. yield increases) offset some of the losses from declining socio-economic capitals. Conversely, in technologically advanced scenarios (e.g. SSP4), where manufactured and financial capitals increase greatly, demands for services could be met relatively easily <u>during most of the century</u>, leading to a decline in intensive management because of a lack of need, rather than a lack of opportunity.

Results also show some broad geographical patterns. While the most unproductive areas of Europe (e.g. mountain ranges, high latitudes) were the most resistant to change under any scenario, other areas responded differently depending on the scenario conditions. South-eastern Europe (Greece, Bulgaria, Romania and Hungary), was slightly more vulnerable to extensification and abandonment where supply levels matched demands as in SSP4 (i.e. when 'benefit' or profit levels were low), but were more robust to low levels of capitals in SSP3. In contrast, Western Europe (particularly Germany, France, England and intensively-managed areas of Spain) suffered widespread abandonment in SSP3. As climate change increased in magnitude

25 through RCPs 2.6, 4.5 and 8.5, land management in North-Eastern Europe (Poland, Lithuania, Latvia, Estonia and southern Finland) tended towards forestry, as increases in forest productivity and decreases in crop productivity made arable agriculture less competitive.

Behavioural parameter variations had distinct effects in different scenarios and on different metrics (Table 2, Figure 2 & Supplement 2Appendix C). In general, land use patterns in scenarios with less intensive management (and also lower land use

30 fragmentation as measured by the fractal dimension) were less affected by behavioural parameter changes; these scenarios included RCP4.5-SSP3, RCP8.5-SSP3 and RCP4.5-SSP4. Conversely, scenarios with more intensive management (RCP4.5-SSP1, RCP2.6-SSP1 and RCP2.6-SSP4) were more affected, producing more fragmented land systems. These differences were not, but not necessarily reflected in different levels of ecosystem service supply levelsies, however (Table 2, Fig. 2). Instead, these showed increasing effects of parameter variations over the simulation period, with the largest effects occurring

in SSPs 3 and 4 and the smallest in SSP1. In two scenarios (RCP2.6-SSP4 and RCP4.5-SSP3), behavioural parameterisations determined whether food was over- or under-supplied by the 2080s. These differences were also partially correlated with climatic scenarios, with the more productive-land systems that were more productive under high-end scenarios proving more robust to behavioural differences (particularly RCP4.5-SSP4, RCP4.5-SSP1 and RCP8.5-SSP5, in which variations had almost

- 5 no impact on food supply). Of the two forms of variation simulated, increased requirements for benefit from land management (thresholds) led to increased fragmentation within scenarios on average, but also increased differences in fragmentation between them. Individual variation increased the differences in fragmentation <u>between scenarios</u> more than the-it did average <u>levels across them</u>, but at the higher strength these differences were reduced to approximately baseline levels, suggesting a 'peak effect' under small levels of individual variation beyond which the most irrational agents were selected outthe effects of
- 10 behavioural parameters were reduced (Fig. C1). Plots of the evolution in behavioural parameter values during the simulations (Fig. C2) show that this effect was partially due to a context-specific 'selection pressure' towards particular parameter values, especially in the SSP3 simulations. In this context, the initially random distribution of agent parameters became skewed towards higher values of competition thresholds, lower values of abandonment thresholds and lower variation between agents, demonstrating a disproportionate persistence of agents who were relatively unlikely to respond to benefit values.
- 15 In all cases, the delicate balance between food and timber production highlights the sensitivity of results to demand levels for ecosystem services derived from agriculture and from forestry. In many cases, simulations resulted in widespread adoption of multifunctional land uses that provide both sets of services to some extent, with the locations of these being scenariodependent. The levels of demand, relative valuation and production of these services therefore appear to be major determinants of the nature of European land systems in this model.

20 4 Discussion

The work presented here highlights the importance of both model design and scenario conditions for understanding possible future change in large-scale land systems. This complements previous findings that model design and initial data conditions had a greater impact than scenarios on simulated land use change (Alexander et al., 2017), but extends the comparison to new design and scenario components. Until now, exploration of these has been generally limited to optimising pattern-based models

- 25 and the biophysical and economic factors that they incorporate, neglecting the social conditions and processes that often vary dramatically between scenarios (Brown et al., 2017; von Lampe et al., 2014; Pedde et al., 2019a). This model implementation demonstrates that agent-based modelling of socio-ecological systems at continental scales is both a feasible and informative method for scenario exploration, producing clear and distinct outcomes that respond directly to scenario definitions. These responses include breakdown of the simulated land system, in which rapid and sub-optimal land
- 30 use changes lead to severe shortages of ecosystem services including food. While such breakdown is occasionally a feature of real-world land systems and a plausible result of severe pressures in the future (Ehrlich and Ehrlich, 2013; Hazell and Wood, 2008; Weiss and Bradley, 2001), it is largely beyond the reach of precluded by equilibrium or optimisation assumptions in

conventional modelling approaches (Balint et al., 2017; Brown et al., 2016a; Farmer and Geanakoplos, 2009). The ability to explore such breakdowns, whether in agent-based or flexible economic models, is clearly necessary for attempts to achieve the converse; stability and sustainability in socio-ecological systems.

To allow proper interpretation, the remainder of this Discussion is divided between technical considerations relating to model design and parameterisation, and reflection on the results produced in this study.

4.1 Model design

CRAFTY-EU is an explorative model, and is not designed to predict (inherently unpredictable) land system changes (Brown et al., 2016a). Further, the CRAFTY framework is intended to provide relatively simple, generic methods for exploring land manager decision-making over large geographical extents (Murray-Rust et al., 2014), and is used here to represent decision-

10 making within local land systems rather than at the level of individual managers. As such, this model application is a first-step towards improved understanding of behavioural processes within large-scale land systems. At a general level, the results presented here are realisations of a single approach to land systems modelling, which complement

alternative projections made by other models (e.g. (Harrison et al., 2019; Stürck et al., 2018; Verkerk et al., 2018)). In particular, conceptual or theoretical frameworks within which behavioural modelling can occur are diverse and disputed, and

- 15 a universally applicable representation of the complex social processes involved in land use change is not available or even, necessarily, possible (Brown et al., 2016a; Huber et al., 2018; Meyfroidt et al., 2018). Even given this caveat, our exploration of behavioural parameters is illustrative rather than exhaustive, intended to reveal the implications of basic assumptions more than exact parameter values. We do not attempt to derive realistic parameter values but instead assess model sensitivity across a large range, from identical agents attempting to maximise their benefit, to highly diverse agents tolerating benefit differences
- 20 of around 100% of mean values (Table A4). Indeed, the CRAFTY framework is designed specifically to allow exploration of abstracted behaviours that do not require precise parameterisation. In this respect, this study we deliberately builds on earlier studies of the parameterisation of behavioural processes in CRAFTY, including in a similar scenario context (e.g. (Brown et al., 2014b, 2018b)). While the CRAFTY framework is designed specifically to allow exploration of abstracted behaviours that do not require precise parameterisation, linking these values to empirical data is clearly desirable even where apparent effects on model outcomes are small.
- In this case, more precise parameterisation might necessitate finer spatial and behavioural resolution, as is typical of many agent-based models. Nevertheless, conceptual and practical limits to behavioural parameterisation exist, and the aggregation of agents by location, land use and behaviour (or combinations such as functional type) is an established approach to large-scale modelling (Arneth et al., 2014; Valbuena et al., 2010). Identifying an appropriate balance between scale and resolution
- 30 in agent-based modelling of the European land system certainly requires further investigation of behavioural processes and data across scales, but the fact that we find clear systemic impacts of behaviours simulated at 10 arcminute resolution implies that finer resolutions may well be informative, as they have proved in previous applications of the CRAFTY framework (e.g.

(Blanco et al., 2017). At present, data availability is the greatest barrier to the adoption of these resolutions in European modelling.

A number of more specific considerations are also important for interpreting our findings. Most significantly, the simulations presented here form an experiment into the effects of simulating land management as the provision of multiple (but arbitrarily

- 5 limited) ecosystem services, which depend upon a set of scenario-dependent capitals and which are valued equally per standardised unit of demand. This design ensures that trade-offs between services are clear, but does not assume preferential production of some services (such as food) when supply levels are equally insufficient. As a result, scenarios in which shortfalls in service provision exist <u>might</u> represent an artificially balanced outcome, <u>with-with</u> real-world equivalents <u>expected to potentially</u> diverginge towards more homogeneous land uses to some extent. In this respect, our findings suggest that further
- 10 exploration of trade-offs between service provision, in terms of both production systems and valuation, should be a priority for land system modelling. This is especially important given potential changes in current valuation practices, for example through carbon pricing or payments for ecosystem services, which could transform the competitiveness of currently minor land uses and require models to account for the services that they produce (Kay et al., 2019).
- Beyond Europe, neither CRAFTY-EU nor the IAP that is used to calibrate it explicitly represent production and trade. While scenario-specific import levels are assumed, these are likely to be overestimates in challenging scenarios with large shortfalls in service provision that imply shortages elsewhere in the world (Dellink et al., 2017; Harrison et al., 2015; Stevanović et al., 2016). This would reduce supply levels even more than simulated here. Furthermore, alternative treatments of international trade based on assumptions of economic equilibrium would be inconsistent with the supra-economic behavioural approach used in CRAFTY-EU (Arthur, 2006). The relative provision of different services is also subject to substantial uncertainty in
- 20 our representation of forest growth, with assumed adaptation to changes in species' suitability likely to overestimate real-world adaptation (Schelhaas et al., 2015), as the CRAFTY framework has previously been used to demonstrate (Blanco et al., 2017b, 2017a).

Notwithstanding the above limits on the model's accuracy, the robust, cross-sectoral nature of the model, building on the established and evaluated IAP and CRAFTY framework, means that it is capable of providing well-founded and novel insight

25 into land system dynamics. Model evaluation performed for this and earlier studies has revealed no clear biases or instabilities, with CRAFTY-EU producing realistic outcomes in the absence of information about baseline land uses. The responses of the model to the scenarios can be seen as coherent responses to a set of land system drivers that are fully interpretable in light of transparent model assumptions.

4.2 Model results

30 A key finding of this work is that the sensitivity of land use to social (as well as economic and climatic) conditions makes land systems vulnerable to breakdown when these conditions worsen substantially. Such worsening is a key characteristic of some future scenarios (e.g. SSP3), but one that has generally only been explored through qualitative scenario descriptions (e.g. (Cradock-Henry et al., 2018; Kebede et al., 2018; Pedde et al., 2019a)). In SSP3, declines in socio-economic capitals are so

precipitate and substantial that the resulting breakdown of the simulated land system is highly plausible, and proves almost impossible to avoid in our modelling, regardless of exact parameterisations. In other model projections of this scenario, similar outcomes are avoided only by very large increases in food prices that compensate for relatively low crop yields and stimulate food production at the expense of forest cover (Doelman et al., 2018; Fujimori et al., 2017; Harrison et al., 2019; Hasegawa et

al., 2015; Popp et al., 2017). <u>These large increases are generally enabled by an assumption of economic equilibrium, and not constrained by financial or other productive limitations related to the scenario.</u>
 This implies that scenario modelling using economic equilibrium assumptions could prove misleading where scenario

conditions place<u>potentially non-equilibrium</u> limits on price or production levels. Substantial declines in financial and manufactured capitals, for instance, may effectively preclude the necessary economic stimuli or production responses to meet

- 10 demand in SSP3. While this problem is starkly illustrated by non-equilibrium modelling such as that presented here, <u>it is not</u> in principle impossible to account for it in models that assume equilibrium dynamics within scenario-specific constraints (Heistermann et al., 2006). <u>However</u>, <u>its</u>-knock-on effects on consumption, demand and supply (and wider socio-economic systems) are obscured by the pre-definition of those factors in scenario storylines. The CRAFTY-EU model therefore makes only one of two crucial connections, linking social conditions and supply levels through the capitals-production relationships
- 15 without completing the link back to demand. Most starkly, if insufficient food is produced to maintain population levels, populations and subsequent demand would inevitably decrease – a fundamental feedback that remains absent from scenario modelling.

Such an effect is particularly pertinent to our simulations of SSP3. In the storylines of SSPs 3 and 4, large populations suffer low and unequally-distributed economic growth, leaving per capita demand for agricultural production approximately half that

- 20 in the wealthier and more equitable contexts of SSPs 1 and 5 (Table 2; (Kok et al., 2018; Riahi et al., 2017). The large and long-lasting shortfalls in food production generated by SSP3 simulations therefore imply far greater challenges than the shortfalls generated by SSP1 simulations, because the latter allows adequate (if undesirable) levels of nutrition to be maintained while the former does not. This is clearly an important area for further research, albeit one that requires much improved treatment of food security and its consequences as well as fully internally consistent scenario simulations that account for a
- 25 range of economic and social factors. Such internal inconsistencies in modelled scenarios are not limited to socio economic systems; scenario assumptions about the magnitude of elimate change are widely made without accompanying assumptions about implied land based mitigation actions (Kriegler et al., 2017). For analyses of future scenarios to be dependable, all of these issues need to be addressed.

In addition to identifying very large negative impacts of some scenario combinations, we also find that these impacts differ

30 widely across Europe. Some areas appear to face high likelihoods of substantial changes; for example we find that Eastern Europe is broadly more vulnerable to changes in demand levels (and hence 'benefit' or price levels) and Western Europe broadly more vulnerable to changes in capital levels. Many of the worst simulated outcomes have notable mirror-images in history, where land systems gradually became more intensive, homogeneous and efficient as financial, technological and social capitals developed (e.g. (Petit and Lambin, 2002)). Projected declines of these capitals produce a return to fragmented,

extensive production in our simulations; a reverse precedent that adds some credibility to model responses, while clearly not suggesting predictive accuracy. It is also notable that greater climatic change can actually ameliorate the worst outcomes in some cases (e.g. SSP3) because it allows higher yields in parts of Europe to offset losses and socio-economic difficulties elsewhere - at least in the absence of the pests, diseases and extreme weather events that may be associated with such climatic

- 5 <u>change</u> (Donatelli et al., 2017). Similarly, technologically advanced scenarios (e.g. SSP4) allow relative ease of production and therefore free-up land, leading to some extensification and abandonment.
- Within these broad findings, variations in behaviour can have substantial effects. These are <u>generally</u> more pronounced in lowend climate change scenarios that rely on slight competitive advantages of intensive land systems to meet service demand levels, and which are therefore sensitive even to slight <u>deviations from simple</u> economic <u>irrationality rationality</u> in management
- 10 decisions. The literature suggests that intensive farmers are more vulnerable to changing price levels (van Vliet et al., 2015), and this vulnerability is amplified here, as it may be in reality, by a reliance on socially-mediated capitals that support farming (Sutherland and Burton, 2011). It is also notable that the behavioural effects we observe are similar at both simulated strengths of behaviour, suggesting that even small differences in land managers' responses to scenario conditions can have substantial consequences. Indeed, increasing_increased_behavioural differences do not necessarily persist in the face of challenging
- 15 scenario conditions, under which some behavioural responses may consistently outperform others. In our SSP3 simulations, such an effect is apparent in the emergence of greater homogeneity and tenacity in the agent population, and the loss of more responsive, economically rational agents (Fig. C2). Wmay lead to the loss of more extreme agents from the system, giving a behavioural saturation effect that could limit the extent of irrationality in the real world. In any case, the idespread evidence already exists of widespread deviation from economically optimal behaviour amongst land managers in the real world, as in
- 20 <u>the, such as</u> selection of economically inferior options for social reasons, or socially-mediated uptake that spans long time periods (Brown et al., 2018a, 2019; Sereke et al., 2016). Our findings further show that the effects of such behaviours on supply levels can accumulate over time (Fig. 2), when permitted to through a path-dependent and non-optimising modelling approach., The inclusion of more realistic and heterogeneous behaviours in land system models therefore appears to be a justifies if not necessitates the incorporation of such behaviour in land systems models. We suggest that this is a pre-
- 25 requisite for accurate assessments of future scenarios, and so for effective land management planning and policy-making.

5 Conclusions

The application of an agent-based model to simulate future European land use change suggests an important role for largescale behavioural models of this kind. CRAF<u>T</u>Y-EU is developed here to investigate broad forms of human behaviour in the context of land management decision-making, and demonstrates that such behaviours can have multiple substantial effects in

30 different scenario contexts. Furthermore, the most notable of these effects were linked to basic model assumptions rather than exact design or parameterisation choices. The inclusion of socio-economic aspects of future scenarios as active drivers of land use decision-making had impacts at least as large as simulated climate change, with behavioural effects further shaping trajectories within those scenarios. Competition between a cross-sectoral, multi-functional range of land uses highlighted the critical importance of the relative valuation of ecosystem services, and the ability of models to represent a relevant range of services. Most prominent, however, was the effect of allowing land use decisions to occur without enforced equilibria or optimisation. In scenarios with challenging socio-economic conditions, this led almost invariably to breakdown of the

- 5 simulated land system, and severe shortages of food and other services. These effects were apparent even at low levels of behavioural complexity, and persisted across tested parameterisations. Indeed, we find some evidence that behavioural effects may be partially 'self-correcting', with <u>more extremesome</u> behaviours being selected out by a competitive process. These findings show a clear need and scope to consider the role of human behaviour in shaping land system development, <u>whether</u> through the development of new large-scale behavioural models or the introduction of additional behavioural and scenario-
- 10 related constraints in economic models. Although this task remains challenging, the data and tools to explore social dimensions of scenario space are developing rapidly, and appear capable of providing important new insights into the future development of large-scale land systems.

Code and data availability

The full model code and date are available for download and visualisation at https://landchange.earth/CRAFTY

15 Appendices

Appendix A: Model parameterisation Appendix B: Model evaluation

Supplements

Supplement 1: Demand files, giving ecosystem service demand values for each scenario and year.

20 Supplement 2: Further graphical results summaries.

Author contribution: CB developed the model and drafted the manuscript; BS & CB performed sensitivity analyses and BS developed calibration routines and the web platform; MR advised on model development and interpretation, and all authors finalised the manuscript.

25 Competing interests

The authors declare that they have no conflict of interest

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Agent functional type	Ecosystem services produced	Area covered (% cells in baseline)
Intensive arable farming	✓ ■ ½ ~ ¾	12.6%
Intensive pastoral farming	****	4.8%
Intensive agro-forestry mosaic	* ~ ** ***	10.8%
Intensive farming	* 134 >> 🖛	5.9%
Managed forestry	\$ [™] → [™]	15.0%
Mixed farming	\$ 1 <u>€</u> * * * *	5.2%
Mixed pastoral farming	***	1.9%
Mixed forest	<u>≷</u> * * *	0.3%
Extensive pastoral farming	*****	0.9%
Extensive agro-forestry mosaic	<i>~</i> ⇒ 🕅 🐂 🖄 ⇒	4.8%
Very extensive pastoral farming	* *	2.3%
Multifunctional	\$ ™ ~ % √ ™	18.3%
Minimal management	<u>الا</u> الا الح	6.5%
Unmanaged land	ざみ	9.7%
Unmanaged forest	13 环 -	0.3%

Peri-urban	* 🕅 👬 🛲 📎	0.7%
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Table 1: Details of Agent Functional Types (AFTs) used in CRAFTY-EU. Ecosystem services are represented as follows: crops:

i; meat: T; timber: i; carbon sequestration: i; diversity: ; recreation: i. Primary ecosystem services of each AFT are those produced in quantities at least 50% of the maximum of any other AFT, and are shown in black. Secondary ecosystem services are those produced in lower quantities, and are shown in grey. The initial distribution of these AFTs across modelled grid cells, full parameterisation of capital sensitivities and production levels are described in full in Appendix A. The conceptualisation and parameterisation of AFTs allows for some variation in capital sensitivities, service production abilities and land uses within each AFT. The Urban (not shown) and Peri-urban AFTs are included only as placeholders for urban modelling in the IAP, and are constrained to reproduce the same results here, with Peri-urban also allowing for surrounding production of other ecosystem

10 services as shown.





Figure 1: Maps of simulated land cover in 2086 under the RCP4.5-SSP1 scenario combination (top lefta) and the RCP4.5-SSP3 scenario combination (top rightb), showing the two extremes of modelled outcomes across the simulated scenarios. These extremes are driven by the radically different socio-economic capital levels within the two scenarios (bottom; capitals shown as mean values, normalised by their initial mean value). Changes in capital trends occur due to their interpolation between discrete IAP data points.





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Figure 2: Total shortfalls in fFood (meat and crops) supply as a proportion of demand production (meat and crops) as proportion of demand levels-in all simulated scenario combinations. Negative values indicate a surplus and positive values Values less than 1 indicate a shortfall. Shaded areas show the ranges of results produced by simulated behavioural variations within each scenario, with the range in RCP8.5-SSP5 being too small to be visible. Absolute per capita demand levels (including non-food usage and waste of food crops) are approximately twice as large by 2080 in SSPs 1 and 5 as in SSPs 3 and 4.

Scenario	Explanation	Main Results	Effects of behavioural
RCP2.6 – SSP1	Represents a future in which limited climate change occurs, and socio-economic conditions gradually improve through economic growth, stable government, high social cohesion and international cooperation. Demand levels reflect widespread over-consumption of food	Gradually increasing shortfalls in supply levels of most services, especially timber, <u>limiting scope for</u> <u>overconsumption</u> , <u>over the</u> <u>simulation period</u> . Intensive management across much of Europe, with more extensive land uses in northern and southern latitudes. Relatively stable AET dynamics	Increased thresholds and individual variations produced more intensive & efficient land uses and more use of unmanaged land, but similar supply levels
RCP2.6 – SSP4	Represents a future in which limited climate change occurs, but large economic inequalities and fluctuations develop and contribute to low social cohesion. Nevertheless, substantial technological investment is made and environmental protection is prioritised.	Broadly increasing service provision in first half of the century, driven by large increases in manufactured and financial capitals, leading to surpluses, especially of meat. Subsequent dramatic drops in intensively managed areas mid-century and tendency to abandonment, minimal management or extensive management, especially away from central Europe; development of shortfalls. Fragmentation of land use.	Increased thresholds and individual variations produced more intensive & efficient land uses particularly in Central- Western Europe, with substantial increase in meat supply and small drop in crops supply <u>that</u> vary widely between parameterisations.
RCP4.5 – SSP1	Represents a future in which low-medium climate change occurs, and socio-economic conditions gradually improve through economic growth, stable government, high social cohesion and international cooperation. Demand levels reflect widespread over-consumption of food.	Relatively stable service supplies but consistent shortfalls in timber production. Widespread intensive management of land, with little change from baseline.	Increased thresholds and individual variations produced more abandonment in Central- Eastern Europe, with more timber production and otherwise similar supply levels.
RCP4.5 – SSP3	Represents a future in which low-medium climate change	A very dynamic scenario in which land uses fluctuate in	Very similar results across all parameterisations, with

	occurs, while social and economic conditions worsen, with limited and ineffective political responses <u>and</u> relatively low levels of food <u>consumption.</u> -	response to rapidly declining capital levels. Very large shortfalls develop, especially of foodsuggesting serious shortages of food, although these are rapidly reduced disappear after 2070. Widespread extensification and abandonment of land occurs across Europe.	effects on food supplies small until the final decade of the simulation period.
RCP4.5 – SSP4	Represents a future in which low-medium climate change occurs, and large economic inequalities and fluctuations develop and contribute to low social cohesion. Nevertheless, substantial technological investment is made and environmental protection is prioritised.	Substantial surpluses are produced thanks to increasing financial and manufactured capitals. Fluctuations in land management result in a changeable and fragmented land system, with extremes of intensive and very extensive land management co-existing in many areas.	Similar results across all parameterisations, with behavioural differences leading to slightly less extensification and slightly larger surpluses.
RCP8.5 – SSP3	Represents a future in which high-end climate change occurs, while social and economic conditions worsen, with limited and ineffective political responses <u>and</u> <u>relatively low levels of food</u> <u>consumption</u> .	As with RCP4.5-SSP3, land management and service supplies are very dynamic, with different trajectories throughout the century, producing large shortfalls that are eventually overturned. Slightly increased average crop productivity supports some intensive management in an otherwise highly fragmented, extensively- managed land system.	More intensive management in Central- Western Europe and more abandonment in Eastern Europe, giving similar service levels with larger surpluses by the end of the period.
RCP8.5 – SSP5	Represents a future in which high-end climate change occurs, while substantial emphasis is placed on social and economic development, fossil fuel exploitation and technology. Demand levels reflect widespread over- consumption of food.	Increases in all capitals allow consistent surpluses of food and timber. Despite a slight general trend towards extensification, most of Europe remains under intensive management.	Very similar results across all parameterisations, with only negligible effects on food supply levels.

Table 2: Identities and characteristics of Representative Concentration Pathway – Shared Socioeconomic Pathway combinations used in CRAFTY-EU simulations presented here. Graphical results are shown in Supplement 2, and full descriptions of the scenarios used can be found in (Kok et al., 2018).

Appendix A: Model parameterisation

This appendix describes the parameterisation of CRAFTY-EU, including the derivation of Agent Functional Types (AFTs). As outlined in the main text, AFT identities were designed to capture important sectoral and cross-sectoral land systems at local (10') scale. The initial distribution of these AFTs across the modelled land surface was then determined on the basis of

- 5 land use categories modelled by the IMPRESSIONS Integrated Assessment Platform (IAP) under baseline conditions (Harrison et al. 2015) (Fig. A1). This distribution ensured a common starting point for the two models that was fully consistent with the capital levels, demand levels and scenario conditions applied here, so that subsequent simulated changes could be attributed to changes in those conditions rather than inconsistencies in calibration data. The mapping of IAP output land use categories to AFTs is described in Table A1.
- 10 Table A1: The composition of Agent Functional Types (AFTs) in *CRAFTY-EU* in terms of baseline IAP land use categories. In any case where the given IAP categories occupy more than 70% of a cell, that cell is allocated to the corresponding AFT in the baseline map of *CRAFTY-EU*, except in the case of the Peri-urban AFT, for which the threshold (of urban area) is 40%.

Agent Functional Type	Composition
Intensive arable farming	Intensively farmed
Intensive pastoral farming	Intensively grass
Intensive agro-forestry mosaic	Intensively farmed, intensively grass, managed forest
Intensive farming	Intensively farmed, intensively grass
Managed forestry	Managed forest
Mixed farming	Intensively farmed, intensively grass, extensively grass
Mixed pastoral farming	intensively grass, extensively grass,
	very extensively grass
Mixed forest	Managed forest, unmanaged forest
Extensive pastoral farming	Extensively grass
Extensive agro-forestry mosaic	extensively grass,
	very extensively grass, managed forest
Very extensive pastoral farming	Very extensively grass
Multifunctional	4 or more land uses in uncommon combination
Minimal management	very extensively grass, unmanaged forest, unmanaged land
Unmanaged land	Unmanaged land
Unmanaged forest	Unmanaged forest
Peri-urban	Any combination with $> 40\%$ urban area
Urban	Urban

The abilities of these AFTs to utilise capitals (Table A2) and produce ecosystem services were defined via capital sensitivity

- 15 and productive ability parameters (given, for each AFT, in Table A3). Where possible, values were derived from the IAP, and so preserved common forms of secondary land management and ecosystem service production within each AFT. Values that had no equivalent in the IAP (e.g. recreation service provision levels) were assumed on the basis of land management intensity and diversity, with variations used to understand the significance of these assumptions. This was also the case with the modelled biodiversity ecosystem service, which was here represented through the proxy of land use diversity (labelled
- 20 Diversity below) within each AFT.

In CRAFTY-EU, modelled production of ecosystem services occurs subject to capital levels, according to the equation

$$p_{s,i,t} = o_{s,t} \prod_{c} c_i^{\lambda_{c,t}}$$
(1)

Where $p_{(s,i,t)}$ represents the level of production of ecosystem service s in cell i by AFT t, calculated as the product across all capitals c of cell-specific capital levels ci weighted by the sensitivity $\lambda_{(c,t)}$ (black rows in tables below) of the AFT t to the capital c, multiplied by the maximum level of production $o_{(s,t)}$ (red rows in tables below) that the AFT is able to produce.

5 Maximum production levels $o_{(s,t)}$ and capital sensitivities $\lambda_{(c,t)}$ are constant throughout simulations, while capital levels c_i vary according to scenario and, potentially, previous production levels and institutional intervention. Maximum production levels can, however, vary across individual agents within AFTs, and do so here, in some experiments, randomly according to Gaussian distributions around the mean value (Tables A3 & A4).

The ability of an AFT to produce a service was first established by checking the average production level of each service across cells assigned to that AFT under baseline conditions. If this average value was greater than or equal to 1% of the largest value produced by any AFT, that service was added to the AFT's productive abilities. The exact AFT-specific maximum production value (o_(s,t;)) was calculated by extracting the 100 most productive cells for AFT t of service s and fitting a Gaussian distribution to the production levels in those cells using the R package fitdistrplus (Delignette-Muller and Dutang 2015). The

15 of random variation in production levels. This procedure was used under the assumption that the 100 most productive cells represented optimal production conditions, and therefore provided a suitable basis to estimate production levels in the effective absence of capital constraints.

mean of this fitted distribution was taken as the value of $o_{(s,t)}$, while the standard deviation was retained for the introduction

Capital levels were derived from outputs of the IAP to provide baseline and scenario-specific values (capitals are defined in Table A2). IAP results were interpolated to provide annual values for each capital on each grid cell within each scenario, for

20 the period 2010-2100. Where the derivation of capital values involves simulated quantities of production, these were normalised by the terrestrial area available in each cell (also an output of the IAP).

AFT-specific capital sensitivities $\lambda_{(c,t)}$ were then estimated by plotting all production levels of service s by AFT t against each capital in turn (e.g. Fig. A1), with relationships quantified between the extremes of linear relationships (which were assigned a sensitivity value of 1.0) and random relationships (which were assigned a sensitivity value of 0.0). This procedure

25 did not, and was not intended to, replicate the land use allocation methods applied in the IAP, but to generate similar sensitivities on the basis of which agent decision-making could proceed.
Once these relationships were established, IAP output maps were used to quantify demand levels for each of the modelled

ecosystem services by calculating service production levels according to the optimal production and capital sensitivity values described above. This was repeated at each timestep (2020s, 2050s and 2080s in the IAP, which were linearly interpolated to

30 annual values between 2016 and 2086 for CRAFTY-EU). Where the IAP projected a shortfall in service production, the supply was calculated and then scaled up to the equivalent of 100% to give a figure for demand. For the services not directly simulated by the IAP (recreation and diversity), the supply levels calculated from IAP output maps were taken as being equal to demand.

These demand levels (given in full in Appendix 3) were then used to calculate context-specific 'benefit' values of production as a basis for competition between agents. Benefit functions were defined to give the value of a certain level of production under a certain level of unmet demand, according to the equation:

 $m_s = u_s(r_s);$

- 5 where m_s is the marginal benefit for service s, u_s is a function that describes the benefit (utility) of production of service s and r_s is the residual demand for service s (Murray-Rust et al. 2014). Linear forms of us were used here, calibrated to ensure equal relative valuation of services; i.e. the production of an equal proportion of unmet demand was assigned an equal benefit value whatever the service. This created a balanced competition between agents that was not skewed towards any particular service(s), with no benefit accruing from production when there was no unmet demand, prompting production under shortfalls
- 10 but not under surpluses.

Table A2: Identities and details of modelled capitals. Exact parallels for some capitals were available in the IAP.

Capital	Explanation	Derivation from IAP
Crop productivity	Natural productivity	Average of simulated productivities for winter wheat, spring
	for crops	wheat, winter barley, spring barley, potatoes, sugar beet, winter
		oilseed rape, spring oilseed rape, maize, forage maize, cotton,
		sunflower and soya
Grassland	Natural productivity	Average of simulated productivities for grass, extensive grass
productivity	for grassland	and permanent grass
Forest	Natural productivity	Potential wood yield
productivity	for forest	
Human capital	Availability of labour	Human capital
Social capital	General level of social	Social capital
	support (cohesion,	
	social networks) for	
	production	
Manufactured	Availability of	Manufactured capital
capital	machinery and	
	infrastructure	
	(including for	
	transportation of	
	goods, where	
	appropriate)	
Financial capital	Economic resources	Financial capital
	supporting production	
Urban capital	Suitability for urban	Percentage urban cover of cell
	development (used to	
	constrain distribution	
	of urban land to	
	follow that modelled	
	by the IAP)	

Tables A3 a-q: Tables showing the sensitivities $\lambda_{c,t}$ of each AFT to capital levels and maximum service production levels $o_{s,t}$ (italics) (Eq. 1). Red-vValues in brackets are the standard deviations of Gaussian distributions used in some simulations to randomly assign production levels to individual agents. Standardised units of production are shown for comparability and consistency with model calculations.

5 a) Intensive arable farming

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0.8	0	0	0	0
Forest prod	0	0	0.7	0	0	0
Grass Prod	1	0	0	0	0	0
Financial	0.9	0.8	0.2	0	0	0.4
Human	1	0.8	0.2	0	0	0.7
Social	0.9	0.9	0.2	0	0	0.3
Manufactured	0.6	0.5	0.1	0	0	0.6
Urban	0	0	0	0	0	0
Production	328 (40)<u>0.41</u>	2280 (158)<u>1</u>		422 (29)<u>0.17</u>		
	<u>(0.05)</u>	<u>(0.07)</u>	0 (0)	<u>(0.01)</u>	0.54<u>61</u> (0.01)	0.1 (0.01)

b) Intensive agro-forestry mosaic

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0.3	0	0	0	0
Forest prod	0.1	0.1	1	0	0	0
Grass Prod	0.3	0	0	0	0.1	0
Financial	0.6	0.7	0.2	0	0	0.4
Human	0.5	0.8	0.1	0	0	0.7
Social	0.5	0.6	0.3	0	0	0.3
Manufactured	0.2	0.2	0.2	0	0	0.6
Urban	0	0	0	0	0	0
Production					0.082	
	316 (44)<u>0.4</u>	811 (82)<u>0.36</u>	59 (12) 0.55	481 (10)<u>0.19</u>	(0.03)<u>0.09</u>	
	<u>(0.06)</u>	<u>(0.04)</u>	<u>(0.11)</u>	<u>(0.004)</u>	<u>(0.03)</u>	0.15 (0.02)

c) Intensive farming

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0.2	0.2	0.1	0	0	0
Forest prod	0	0	0.6	0	0	0
Grass Prod	0	0.1	0.1	0.2	0	0
Financial	0.2	0.2	0.4	0.3	0	0.4
Human	0.2	0.2	0.3	0.3	0	0.7
Social	0.1	0.1	0.3	0.2	0	0.3
Manufactured	0.1	0	0.2	0	0	0.6
Urban	0	0	0	0	0	0
Production					0.75	
	715 (75)<u>0.89</u>	1064 (74)<u>0.47</u>	15 (2)<u>0.14</u>	466 (14)<u>0.19</u>	(0.03)<u>0.84</u>	
	<u>(0.09)</u>	<u>(0.03)</u>	<u>(0.02)</u>	<u>(0.01)</u>	<u>(0.03)</u>	0.15 (0.02)

d) Managed forest

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0.1	0	0	0	0
Forest prod	0.2	0	1	0.1	0	0
Grass Prod	0.3	0	0	0	0	0
Financial	0.3	0.1	0.2	0.1	0	0.4
Human	0.1	0.2	0.2	0.1	0	0.7
Social	0.2	0.2	0.2	0.1	0	0.3
Manufactured	0.1	0.1	0.3	0	0	0.6
Urban	0	0	0	0	0	0
Production				2412	0.51	
			108 (18) 1	(117) 0.96	(0.02) 0.57	
	0 (0)	0 (0)	<u>(0.17)</u>	<u>(0.05)</u>	<u>(0.02)</u>	0.5 (0.2)

10 e) Extensive pastoral farming

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0.4	0	0	0	0	0
Forest prod	0	0	0.7	0	0	0
Grass Prod	0.2	0	0.4	0	0	0
Financial	0.1	0.1	0.2	0	0	0.4
Human	0.2	0	0.2	0	0	0.7
Social	0.2	0.1	0.1	0	0	0.3
Manufactured	0.2	0	0	0	0	0.6
Urban	0	0	0	0	0	0
Production					0.44	
	59 (45)<u>0.07</u>		5 (6)<u>0.05</u>	403 (59)<u>0.16</u>	(0.04)<u>0.49</u>	
	<u>(0.05)</u>	0 (0)	<u>(0.06)</u>	<u>(0.02)</u>	<u>(0.04)</u>	0.7 (0.1)

f) Extensive agro-forestry mosaic

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0.1	0.3	0	0	0	0
Forest prod	0	0	1	0	0	0
Grass Prod	0.1	0.2	0.4	0	0	0
Financial	0.2	0.1	0.3	0	0	0.4
Human	0.2	0.1	0.3	0	0	0.7
Social	0.2	0.1	0.3	0	0	0.3
Manufactured	0.1	0	0.3	0	0	0.6
Urban	0	0	0	0	0	0
Production					0.72	
	<u>0.13 (0.04)</u> 105		57 (17)<u>0.53</u>	634 (395)<u>0.25</u>	(0.03)<u>0.81</u>	
	(35)	0 (0)	<u>(0.16)</u>	<u>(0.16)</u>	<u>(0.03)</u>	0.7 (0.1)

g) Multifunctional

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0.1	0	0	0	0
Forest prod	0	0	0.9	0	0	0
Grass Prod	0.7	0.1	0.1	0	0	0
Financial	0.7	0.3	0.2	0	0	0.4
Human	0.4	0.2	0.1	0	0	0.7
Social	0.5	0.3	0.1	0	0	0.3
Manufactured	0	0.2	0.2	0	0	0.6
Urban	0	0	0	0	0	0
Production				2232		
	<u>0.49 (0.06)</u> 388	774 (132)<u>0.34</u>	62 (12) 0.57	(353) 0.89	0.89 (0.02)<u>1</u>	
	(50)	<u>(0.06)</u>	<u>(0.11)</u>	<u>(0.14)</u>	<u>(0.02)</u>	0.5 (0.1)

h) Unmanaged forest

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0	0	0	0	0
Forest prod	0	0	0	0	0	0
Grass Prod	0	0	0	0	0.3	0
Financial	0	0	0	0	0	0.4
Human	0	0	0	0	0	0.7
Social	0	0	0	0	0	0.3
Manufactured	0	0	0	0	0	0.6
Urban	0	0	0	0	0	0
Production				193	<u>0.57</u>	
				(94) 0.08	<u>(0.02)</u> 0.51	
	0 (0)	0 (0)	0 (0)	<u>(0.04)</u>	(0.02)	1 (0.1)

5

i) Intensive pastoral farming

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0.2	0	0	0	0	0
Forest prod	0	0	0	0	0	0
Grass Prod	1	0	0	0	0	0
Financial	0.8	0	0	0	0	0.4
Human	0.6	0	0	0	0	0.7
Social	0.7	0	0	0	0	0.3
Manufactured	0.1	0	0	0	0	0.6
Urban	0	0	0	0	0	0
Production					0.51	
	<u>1 (0.09)</u> 799			513 (57)<u>0.2</u>	(0.03) 0.57	
	(72)	0 (0)	0 (0)	<u>(0.02)</u>	<u>(0.03)</u>	0.1 (0.01)

j) Mixed farming

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0.7	0.5	0	0	0	0
Forest prod	0.1	0	0.8	0	0	0
Grass Prod	0.2	0.2	0.1	0	0.2	0

Financial	0.4	0.2	0.1	0	0	0.4
Human	0.3	0.2	0.2	0	0	0.7
Social	0.4	0.3	0.2	0	0	0.3
Manufactured	0.2	0.1	0.2	0	0	0.6
Urban	0	0	0	0	0	0
Production					0.84	
	<u>0.58 (0.07)</u> 461	922 (132)<u>0.4</u>	14 (4)<u>0.13</u>	401 (35)<u>0.16</u>	(0.03)<u>0.94</u>	
	(54)	<u>(0.06)</u>	<u>(0.04)</u>	<u>(0.01)</u>	<u>(0.03)</u>	0.2 (0.02)

k) Peri-urban

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0.3	0	0	0	0
Forest prod	0.1	0	0.8	0	0	0
Grass Prod	0.6	0.3	0	0	0	0
Financial	0.5	0.1	0.4	0	0	0.4
Human	0.4	0.1	0.2	0	0	0.7
Social	0.3	0.1	0.2	0	0	0.3
Manufactured	0.1	0.1	0.3	0	0	0.6
Urban	1	1	1	1	1	1
Production					0.64	
	<u>0.11 (0.08)</u> 86	143 (161)<u>0.06</u>	9 (9)<u>0.08</u>	404 (64)<u>0.16</u>	(0.07)<u>0.72</u>	
	(62)	<u>(0.07)</u>	<u>(0.08)</u>	<u>(0.03)</u>	<u>(0.08)</u>	0.2 (0.02)

l) Minimal management

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0	0	0	0	0
Forest prod	0	0	0	0	0	0
Grass Prod	0	0	0	0	0	0
Financial	0	0	0	0	0	0.4
Human	0	0	0	0	0.2	0.7
Social	0	0	0	0	0	0.3
Manufactured	0	0	0	0	0	0.6
Urban	0	0	0	0	0	0
Production				<u>0.17</u>	0.67	
				<u>(0.005)</u> 420	(0.03) 0.75	
	0 (0)	0 (0)	0 (0)	(13)	<u>(0.03)</u>	1 (0.1)

5

m) Mixed pastoral

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0.2	0	0	0	0	0
Forest prod	0	0	0.7	0.1	0	0
Grass Prod	0.6	0	0.3	0	0	0
Financial	0.6	0	0.3	0	0	0.4
Human	0.6	0	0.2	0	0	0.7
Social	0.6	0	0.2	0	0	0.3
Manufactured	0.2	0	0.2	0	0	0.6
Urban	0	0	0	0	0	0

Production	<u>0.61 (0.1)</u> 484		9 (6)<u>0.08</u>	4 91 (67)<u>0.2</u>	0.7 (0.04)<u>0.79</u>	
	(76)	0 (0)	<u>(0.06)</u>	<u>(0.03)</u>	<u>(0.04)</u>	0.35 (0.1)

n) Unmanaged land

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0	0	0	0	0
Forest prod	0	0	0	0.7	0	0
Grass Prod	0	0	0	0	0	0
Financial	0	0	0	0	0	0.4
Human	0	0	0	0	0	0.7
Social	0	0	0	0	0	0.3
Manufactured	0	0	0	0	0	0.6
Urban	0	0	0	0	0	0
Production					0.38	
				<u>1 (0.1)</u> 2515	(0.05) 0.43	
	0 (0)	0 (0)	0 (0)	(254)	<u>(0.06)</u>	1 (0.1)

o) Urban (produces only urban area to replicate that simulated by the IAP)

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0	0	0	0	0
Forest prod	0	0	0	0	0	0
Grass Prod	0	0	0	0	0	0
Financial	0	0	0	0	0	0
Human	0	0	0	0	0	0
Social	0	0	0	0	0	0
Manufactured	0	0	0	0	0	0
Urban	0	0	0	0	0	0
Production	0(0)	0 (0)	0 (0)	0(0)	0 (0)	0 (0)

5

p) Mixed forest

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0	0	0	0	0
Forest prod	0	0	1	0	0	0
Grass Prod	0	0	0	0	0	0
Financial	0	0	0.3	0	0.2	0.4
Human	0	0	0.4	0	0	0.7
Social	0	0	0.1	0	0	0.3
Manufactured	0	0	0.1	0	0.1	0.6
Urban	0	0	0	0	0	0
Production					0.51	
			61 (14) 0.56	356 (69)<u>0.14</u>	(0.03)<u>0.57</u>	
	0 (0)	0 (0)	<u>(0.13)</u>	<u>(0.03)</u>	<u>(0.03)</u>	1 (0.1)

q) Very extensive pastoral

	Meat	Crops	Timber	Carbon	Diversity	Recreation
Crop Prod	0	0	0	0	0	0
Forest prod	0	0	0	0	0	0

Grass Prod	0	0	0	0	0	0
Financial	0	0	0	0	0.1	0.4
Human	0	0	0	0	0.2	0.7
Social	0	0	0	0	0.2	0.3
Manufactured	0	0	0	0	0.1	0.6
Urban	0	0	0	0	0	0
Production	0 (0)	0 (0)	0 (0)	0 (0)	0.04 (0.03)	1 (0.1)



Figure A1: Example capital-service relationships in IAP output data, used to quantify the capital sensitivities for AFTs in *CRAFTY-EU*. Timber production by Mixed Forest agents (a) is found to be almost completely insensitive to grassland productivity capital, giving a $\lambda_{c,t}$ value of 0, while timber production by Managed Forest agents (b) is highly sensitive to forest productivity, giving a $\lambda_{c,t}$ value of 1.

Table A4: Behavioural parameter variations used in the simulations. Parameter set 1 is the default from which main results are derived; in this setup agents respond directly to benefit values with no additional individual or typological behaviour. In parameter set 2, giving upabandonment -and giving incompetition thresholds are altered to introduce abandonment of land when benefit values fall below the giving upabandonment threshold value, and resistance to change unless a competing land use has an additional benefit value of at least the giving incompetition threshold. Intensive land use agents are parameterised to be less tolerant of low benefit values, and more willing to switch to a land use with higher benefit values. In parameter set 3, individual agents differ from one another in terms of their abilities to produce different ecosystem services, and their giving upabandonment and giving incompetition thresholds. Parameter sets 4 and 5 replicate parameter sets 2 and 3 respectively, but with larger values for thresholds and variations, with threshold values of up to approximately 100% of mean benefit values.⁻ Abbreviated AFT names are as follows: EP = Extensive Pastoral, Ext_AF = Extensive agro-forestry, IA = Intensive agro-forestry, Int_Fa = Intensive farming, IP = Intensive pastoral, MF = Managed forest, Min_man = Minimal management, Mix_Fa = Mixed farming, Mix_For = Mixed forest, Mix_P = Mixed pastoral, Multifun = multifunctional, P. Ur = Peri urban, UL = Unmanaged land, UMF = Unmanaged forest, Ur = Urban, VEP = Very extensive pastoral.

Param-set 1

(Behavioural baseline)

		Competition threshold	Abandonment	Abandonment threshold			
	givingInDistributionMe	<u>(standard</u>	<u>threshold</u>	<u>(standard</u>	serviceLevelNoiseMinMi	serviceLevelNoiseMaxM	givingUpProbAbando
	an <u>Competition</u>	deviation)givingInDistribut	<u>(mean)givingUpDis</u>	deviation)givingUpDistributi	<u>nimum service</u>	aximum service	<u>nment probability</u>
Name	<u>threshold (mean)</u>	ionSD	tributionMean	onSD	production (relative)	production (relative)	<u>(per timestep)</u>
E <u>xtensive</u>							
<u>Pastoral</u> P	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Ext <u>_AFensive</u>							
<u>Agroforestry</u>	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Intensive Arable	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Intensive							
Agroforestry	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Intensive Farming	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Intensive Pastoral	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Managed Forest	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Minimal							
Management	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Mixed Farming	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Mixed Forest	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Mixed Pastoral	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Multifunctional	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Peri-Urban	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Unmanaged Land	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Unmanaged							
Forest	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000
Urban	0.00000	0.00000	100.00000	0.00000	1.00000	1.00000	0.00000
Very Extensive							
Pastoral	0.00000	0.00000	0.00000	0.00000	1.00000	1.00000	0.00000

(Increased thresholds)							
			Abandonme nt				Abandonment
Name	Competition threshold	Competition threshold	threshold (mean)	Abandonment threshold (standard deviation)	Minimum service	Maximum service	probability (per timesten)
Extensive	(mean)		(incuri)				unescop
Pastoral	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Extensive							
Agroforestry	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Intensive							
Arable	0.00020	0.00000	0.00030	0.00000	1.00000	1.00000	0.25000
Intensive							
Agroforestry	0.00020	0.00000	0.00030	0.00000	1.00000	1.00000	0.25000
Intensive	0.00000			0.00000	1 00000	4 00000	0.05000
Farming	0.00020	0.00000	0.00030	0.00000	1.00000	1.00000	0.25000
Pastoral	0.00020	0.00000	0 00020	0 00000	1 00000	1 00000	0.25000
Managed	0.00020	0.00000	0.00030	0.00000	1.00000	1.00000	0.25000
Forest	0.00020	0.00000	0.00030	0.0000	1.00000	1.00000	0.25000
Minimal							
Management	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Mixed							
Farming	0.00020	0.00000	0.00030	0.00000	1.00000	1.00000	0.25000
Mixed Forest	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Mixed							
Pastoral	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Multifunction	0.00050	0.00000	0.00010	0.00000	1 00000	1 00000	0 10000
al Dori Urbon	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Linmanaged	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Land	0.00050	0.00000	0.00010	0.0000	1.00000	1.00000	0.10000
Unmanaged	0100000	0.00000	0100010		2100000	2100000	0120000
Forest	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000
Urban	0.00050	0.00000	100.00000	0.00000	1.00000	1.00000	0.10000
Very							
Extensive							
Pastoral	0.00050	0.00000	0.00010	0.00000	1.00000	1.00000	0.10000

Paramset 2

Competition threshold (mean)	Competition threshold (standard deviation)	Abandonment threshold (mean)	Abandonment threshold (standard deviation)	Minimum service production (relative)	Maximum service production (relative)	Abandonment probability (per timestep)	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
0.00020	0.00001	0.00030	0.00002	0.95000	1.05000	0.25000	
0.00020	0.00001	0.00030	0.00002	0.95000	1.05000	0.25000	
0.00020	0.00001	0.00030	0.00002	0.95000	1.05000	0.25000	
0.00020	0.00001	0.00030	0.00002	0.95000	1.05000	0.25000	
0.00020	0.00001	0.00030	0.00002	0.95000	1.05000	0.25000	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
0.00020	0.00001	0.00030	0.00002	0.95000	1.05000	0.25000	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
0.00050	0.00001	0.00010	0.00001	1.00000	1.00000	0.10000	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
0.00050	0.00001	0.00010	0.00001	0 90000	1 10000	0 10000	
0.00050	0.00001	100.00000	5.00000	1.00000	1.00000	0.10000	
0.00050	0.00001	0.00010	0.00001	0.90000	1.10000	0.10000	
	Competition threshold (mean) 0.00050 0.00050 0.00020 0.00020 0.00020 0.00020 0.00050 0.00050 0.00050 0.00050 0.00050 0.00050 0.00050	Competition threshold (mean) Competition threshold (standard deviation) 0.00050 0.00001 0.00050 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00020 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 0.00050 0.00001 <td>Competition threshold (mean) Competition threshold (standard deviation) Abandonment threshold (mean) 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010 0.00050 0.00001 0.00030 0.00020 0.00001 0.00030 0.00020 0.00001 0.00030 0.00020 0.00001 0.00030 0.00020 0.00001 0.00030 0.00020 0.00001 0.00030 0.00020 0.00001 0.00030 0.00050 0.00001 0.00030 0.00050 0.00001 0.00030 0.00050 0.00001 0.00030 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010 0.00050 0.00001 0.00010</td> <td>Competition threshold (mean) Competition threshold (standard deviation) Abandonment threshold (mean) Abandonment (standard deviation) 0.00050 0.00001 0.00010 0.00001 0.00050 0.00001 0.00010 0.00001 0.00020 0.00001 0.00030 0.00002 0.00020 0.00001 0.00030 0.00002 0.00020 0.00001 0.00030 0.00002 0.00020 0.00001 0.00030 0.00002 0.00020 0.00001 0.00030 0.00002 0.00020 0.00001 0.00030 0.00002 0.00020 0.00001 0.00030 0.00001 0.00050 0.00001 0.00010 0.00001 0.00050 0.00001 0.00010 0.00001 0.00050 0.00001 0.00010 0.00001 0.00050 0.00001 0.00010 0.00001 0.00050 0.00001 0.00010 0.00001 0.00050 0.00001 0.00010 0.00001 0.00050</td> <td>Competition threshold (mean) Competition threshold (standard deviation) Abandonment threshold (mean) Abandonment (standard deviation) Minimum service production (relative) 0.00050 0.00001 0.00010 0.00001 0.90000 0.00050 0.00001 0.00010 0.00001 0.90000 0.00050 0.00001 0.00010 0.00002 0.95000 0.00020 0.00001 0.00030 0.00002 0.95000 0.00020 0.00001 0.00030 0.00002 0.95000 0.00020 0.00001 0.00030 0.00002 0.95000 0.00020 0.00001 0.00030 0.00002 0.95000 0.00020 0.00001 0.00030 0.00002 0.95000 0.00050 0.00001 0.00010 0.00001 0.90000 0.00050 0.00001 0.00010 0.00001 0.90000 0.00050 0.00001 0.00010 0.00001 0.90000 0.00050 0.00001 0.00010 0.00001 0.90000 0.00050</td> <td>Competition threshold (mean) Abandonment threshold (mean) Abandonment threshold (mean) Minimum service production (relative) Maximum service production (relative) 0.00050 0.00001 0.00010 0.00001 0.90000 1.10000 0.00050 0.00001 0.00010 0.90000 1.10000 0.00050 0.00001 0.00010 0.90000 1.0000 0.00020 0.00001 0.00030 0.00002 0.95000 1.05000 0.00020 0.00001 0.00030 0.00002 0.95000 1.05000 0.00020 0.00001 0.00030 0.00002 0.95000 1.05000 0.00020 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Paramset 4							
(Larger							
Thresholds)							
	Competition threshold	Competition threshold	Abandonment threshold	Abandonment threshold	Minimum service	Maximum service	Abandonment probability (per
Name	(mean)	(standard deviation)	(mean)	(standard deviation)	production (relative)	production (relative)	timestep)
Extensive							
Pastoral	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Extensive							
Agroforestry	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Intensive							
Arable	0.00040	0.00000	0.00100	0.00000	1.00000	1.00000	0.25000
Intensive							
Agroforestry	0.00040	0.00000	0.00100	0.00000	1.00000	1.00000	0.25000
Intensive							
Farming	0.00040	0.00000	0.00100	0.00000	1.00000	1.00000	0.25000
Intensive							
Pastoral	0.00040	0.00000	0.00100	0.00000	1.00000	1.00000	0.25000
Managed							
Forest	0.00040	0.00000	0.00100	0.00000	1.00000	1.00000	0.25000
Minimal							
Management	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Mixed							
Farming	0.00040	0.00000	0.00100	0.00000	1.00000	1.00000	0.25000
Mixed Forest	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Mixed							
Pastoral	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Multifunction							
al	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Peri-Urban	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Unmanaged							
Land	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Unmanaged							
Forest	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000
Urban	0.00100	0.00000	100.00000	0.00000	1.00000	1.00000	0.10000
Very							
Extensive							
Pastoral	0.00100	0.00000	0.00030	0.00000	1.00000	1.00000	0.10000

Paramset 5 (Larger Variations)							
News	Competition threshold	Competition threshold	Abandonment threshold	Abandonment threshold	Minimum service	Maximum service	Abandonment probability (per
Name	(mean)	(standard deviation)	(mean)	(standard deviation)	production (relative)	production (relative)	timestep)
Extensive Pastoral Extensive	0.00050	0.00010	0.00010	0.00010	0.85000	1.15000	0.10000
Agroforestry	0.00050	0.00010	0.00010	0.00010	0.85000	1.15000	0.10000
Arable	0.00020	0.00010	0.00030	0.00020	0.90000	1.10000	0.25000
Agroforestry	0.00020	0.00010	0.00030	0.00020	0.90000	1.10000	0.25000
Farming	0.00020	0.00010	0.00030	0.00020	0.90000	1.10000	0.25000
Pastoral Managed	0.00020	0.00010	0.00030	0.00020	0.90000	1.10000	0.25000
Forest Minimal	0.00020	0.00010	0.00030	0.00020	0.90000	1.10000	0.25000
Management Mixed	0.00050	0.00010	0.00010	0.00010	0.85000	1.15000	0.10000
Farming	0.00020	0.00010	0.00030	0.00020	0.90000	1.10000	0.25000
Mixed Forest Mixed	0.00050	0.00010	0.00010	0.00010	0.85000	1.15000	0.10000
Pastoral Multifunction	0.00050	0.00010	0.00010	0.00010	0.85000	1.15000	0.10000
al Peri-Urban Unmanaged	0.00050 0.00050	0.00010 0.00010	0.00010 0.00010	0.00010 0.00010	0.85000 1.00000	1.15000 1.00000	0.10000 0.10000
Land Unmanaged	0.00050	0.00010	0.00010	0.00010	0.85000	1.15000	0.10000
Forest Urban Very	0.00050 0.00050	0.00010 0.00010	0.00010 100.00000	0.00010 0.00010	0.85000 1.00000	1.15000 1.00000	0.10000 0.10000
Extensive Pastoral	0.00050	0.00010	0.00010	0.00010	0.85000	1.15000	0.10000



Figure A2: Baseline *CRAFTY-EU* land cover from which all main simulations begin. This baseline map is derived from that of the IAP, which is a modelled land use allocation on the basis of 1961-1990 average climatic conditions and 2010 socio-economic conditions.

Appendix B: Model evaluation

Evaluation of CRAFTY-EU builds on previous evaluations of the agent-based modelling framework from which CRAFTY-EU is implemented, as well as evaluation of previous comparable implementations. These evaluations have included sensitivity and uncertainty analyses (Arneth, Brown and Rounsevell, 2014; Brown, Murray-Rust, et al., 2014; Murray-Rust et al., 2014a;

- 5 Brown et al., 2016; Holzhauer, Brown and Rounsevell, 2018), model inter-comparison (Alexander et al., 2017; I. Holman et al., 2017), and validation against independently simulated and empirical data (V. Blanco, Brown, et al., 2017; V. Blanco, Holzhauer, et al., 2017). These evaluations are wholly or partially relevant to CRAFTY-EU as they deal, at least in part, with the basic architecture and parameters of the modelling framework, which are shared between all applications. The model is also fully open access, with code and (ODD+) descriptions of previous versions published ((Murray-Rust et al., 2014b; V.
- 10 Blanco, Holzhauer, et al., 2017; Holzhauer, Brown and Rounsevell, 2018)), and with CRAFTY-EU itself available in full (code base) or for immediate use (interactive mode) online (from https://bitbucket.org/geoslurg/crafty_cobra_impressions_kit/ and https://landchange.imk-ifu.kit.edu/CRAFTY). Furthermore, input data has been independently verified and evaluated during the development of the IMPRESSIONS IAP, from which CRAFTY-EU is calibrated (Harrison et al., 2012; Brown, Brown, et al., 2014; Kebede et al., 2015; I. P. Holman et al., 2017).
- Evaluation here, therefore, focuses on the specific European implementation of CRAFTY. As described in the main text, evaluation comprised two main exercises involving runs under static, baseline conditions, the first starting from an unassigned (empty) land use map and the second from the baseline land use map derived from the IAP. The purpose of these two exercises was, respectively 1) to check whether baseline conditions would generate a 'realistic' land use configuration purely on the basis of capital levels and AFT characteristics (i.e. in the absence of any spatial information about land management), and 2) to check for divergence in outcomes from a common starting point consistent with other scenario runs.

The first exercise was conducted ten times to check the magnitude of stochastic variation in model outputs, and was expected to produce more variable outcomes for two reasons. Firstly, a number of potential 'solutions' exist to the problem of producing given levels of ecosystem services from a given landscape, and while reality represents one of these, models unconstrained by initial land use maps should be able to produce – and potentially transition between - many others. This is particularly likely

- 25 here given the dependencies of simulated land use decisions on several different factors (multiple capitals, demand levels, and competition between agents). Furthermore, CRAFTY is a non-optimising and stochastic modelling framework with path-dependencies in outcomes, allowing individual simulations to diverge where initial conditions are unstable, as is the case here. Nevertheless, the degree of conformance in general characteristics of these simulations illuminates an important aspect of model stability, as well as revealing the predictability of model responses to aspatial input conditions.
- 30 The second exercise was simpler to interpret, with large differences in land use between the start and end of the simulation taken to indicate model instability under static conditions. Systematic changes would suggest an inconsistency between CRAFTY-EU parameterisation and baseline conditions, and random change would suggest a more general instability. Either of these would also suggest an innate bias in model outputs with the potential to obscure the impacts of simulated scenarios. Model outputs were therefore assessed in terms of the number of agents within each AFT over time.
- 35 The first exercise was initially performed over 800 timesteps, with 20% of cells being randomly selected for potential change. This long timespan and high rate of competition were chosen to exaggerate model dynamics, ensuring that they could be easily assessed through model outputs. Plots of AFT numbers and service levels were checked visually and statistically for stationarity (using Box-Ljung tests for temporal autocorrelation; (Ljung and Box, 1978))). Once an appropriate simulation duration had been identified, 10 further independent simulations were run to this point using different random number
- 40 generator seed values. The outputs of these simulations were then compared in terms of total numbers of agents within each AFT, total service production levels, the spatial consistency of aggregated AFT classes across the ten simulations, and the similarity of these spatial patterns to that in the independent baseline map (to check for spontaneous convergence, which would suggest a broadly 'realistic' response to initial conditions). The second exercise was performed over 100 timesteps, again with a 20% rate of cell selection. This exercise was designed to run in a sufficient number of replicates to identify and understand
- 45 any divergence from stationarity in terms of numbers of agent per AFT, with stationarity again checked for both visually and statistically, and further runs used only where non-stationarity was detected.

Evaluation against historical data were not performed due to the lack of comprehensive data describing capital levels, demand levels and land use maps, other than those produced by alternative models (e.g. (Fuchs et al., 2015)).

Results

1st evaluation exercise

5 The first evaluation exercise did not result in stationarity during the 800-timstep run period (as confirmed by Box-Ljung tests, in which most AFT timeseries had p-values < 0.01 throughout the simulation). This suggests a tendency for ongoing oscillations in agent numbers (and hence service levels). Nevertheless, after an initial period of rapid change, all AFT numbers remained broadly consistent over time, with remaining short-term and apparent long-term fluctuations being small in comparison to overall agent numbers (Fig. B1). A cut-off of 300 timesteps (equivalent to year 2300 in the simulations) was chosen for further analysis, as AFT numbers had achieved representative values by this point.</p>

The numbers of agents belonging to each AFT at the 300th timestep of each of the ten replicate simulations was very similar (Fig. B2a), as were the service levels produced (Fig. B2b). Furthermore, the spatial consistency of aggregated AFT classes was high, and locations frequently agreed with those in the independent baseline land use map (Fig. B3). Aggregated land use classes were used here to check the assignment of land uses rather than specific agent types, which, being considerably more

15 numerous and less discrete, speak to a different aspect of model behaviour (the balance between competitive and productive behaviours of different AFTs, rather than the appropriateness of ecosystem service production in particular locations under given demand levels).



Figure B1: Numbers of agents belonging to each Agent Functional Type throughout an 800-timstep simulation to check for stationarity.





Figure B2: Numbers of agents belonging to each Agent Functional Type at the 300th timestep of each of the ten independent simulations with no initial land use map (a), and service levels as a proportion of demand levels at the same points (b). Figure B2: Numbers of agents belonging to each Agent Functional Type at the 300th timestep of each of the ten independent simulations with no initial land use map (a), and service levels as a proportion of demand levels at the same points (b).



Figure B3: Map of aggregated simulated land covers across the ten evaluation simulations initialised with no baseline land use map. Baseline land covers are shown on the map, with opacity scaled to show the number of evaluation simulations in which that land cover occurred at the 300th timestep.

2nd evaluation exercise

The second evaluation exercise (running the model under baseline conditions starting from the baseline land use map) showed stationarity throughout the simulation period (Fig. B4), and this was confirmed by Box-Ljung tests that showed no evidence of dependence in the timeseries of any of the AFTs. Absolute numbers of agents remained within 15 of the initial number in all cases. This was taken to demonstrate stability in the initial configuration of *CRAFTY-EU*, implying that changes observed during scenario simulations were fully attributable to the parameterisation of those scenarios rather than inherent variability or trends in model dynamics.



Figure B4: Numbers of agents belonging to each Agent Functional Type throughout the 'baseline' run, in which *CRAFTY-EU* was initialised with the baseline land use map and run under static conditions.

Appendix C: Additional behavioural parameter variation results

The behavioural parameter variations explored in this study (and shown in Table A4) were analysed through three main outputs: their effect on food supply levels (Fig. 2), their effect on land use fragmentation (Fig. C1 below) and the emergent changes in those parameter values during simulations (Fig. C2 below).

5 Fragmentation varied broadly between parameterisations as shown in Figure C1, while behavioural parameter values themselves varied systematically in some scenarios (Fig. C2). This was especially the case in SSP3, where the mean abandonment threshold decreased during simulations, the mean competition threshold increased, and the standard deviation of both decreased, indicating a more similar and persistent population of agents than was present at the start of the simulations.



10 Figure C1: Mean fractal dimension in each of the 5 behavioural parameter sets ('Params' 1-5), measuring the fragmentation of land uses in 2086.





Scenario

	RCP2_6-SSP1	
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- ---- RCP2_6-SSP4
- --- RCP4_5-SSP1
- -- RCP4_5-SSP3
- ···· RCP4_5-SSP4
- --- RCP8_5-SSP3
- -- RCP8_5-SSP5

Parameter set

- Paramset1
- Paramset2
- Paramset3
- Paramset4
- Paramset5

(a)

- Paramset1 Paramset2

> Paramset3 Paramset4

Paramset5



Figure C2: Behavioural parameter changes during scenario simulations: abandonment threshold means and standard deviations (a) and competition threshold means and standard deviations (b).

