

Scotland's Rural College

Key challenges and priorities for modelling European grasslands under climate change

Kipling, RP; Virkajarvi, P; Breitsameter, L; Curnel, Y; De Swaef, T; Gustavsson, A-M; Hennart, S; Hoglind, M; Jarvenranta, K; Minet, J; Nendel, C; Persson, T; Picon-Cochard, C; Rolinski, S; Sandars, DL; Scollan, ND; Sebek, L; Seddaiu, G; Topp, CFE; Twardy, S; Van Middelkoop, J; Wu, L; Bellocchi, G

Published in:

Science of the Total Environment

DOI:

[10.1016/j.scitotenv.2016.05.144](https://doi.org/10.1016/j.scitotenv.2016.05.144)

First published: 31/05/2016

Document Version

Peer reviewed version

[Link to publication](#)

Citation for published version (APA):

Kipling, RP., Virkajarvi, P., Breitsameter, L., Curnel, Y., De Swaef, T., Gustavsson, A-M., Hennart, S., Hoglind, M., Jarvenranta, K., Minet, J., Nendel, C., Persson, T., Picon-Cochard, C., Rolinski, S., Sandars, DL., Scollan, ND., Sebek, L., Seddaiu, G., Topp, CFE., ... Bellocchi, G. (2016). Key challenges and priorities for modelling European grasslands under climate change. *Science of the Total Environment*, 566–567, 4 - 0. <https://doi.org/10.1016/j.scitotenv.2016.05.144>

General rights

Copyright and moral rights for the publications made accessible in the public portal are retained by the authors and/or other copyright owners and it is a condition of accessing publications that users recognise and abide by the legal requirements associated with these rights.

- Users may download and print one copy of any publication from the public portal for the purpose of private study or research.
- You may not further distribute the material or use it for any profit-making activity or commercial gain
- You may freely distribute the URL identifying the publication in the public portal ?

Take down policy

If you believe that this document breaches copyright please contact us providing details, and we will remove access to the work immediately and investigate your claim.

1 **Title:** Key challenges and priorities for modelling European grasslands under climate change

2 **Authors:** Kipling, Richard P.^{a*}, Virkajärvi, Perttu^b, Breitsameter, Laura^c, Curnel, Yannick^d, De Swaef,
3 Tom^e, Gustavsson, Anne-Maj^f, Hennart, Sylvain^d, Höglind, Mats^g, Järvenranta, Kirsi^b, Minet, Julien^h,
4 Nendel, Claasⁱ, Persson, Tomas^g, Picon-Cochard, Catherine^j, Rolinski, Susanne^k, Sandars, Daniel L.^l,
5 Scollan, Nigel D.^a, Sebek, Leon^m, Seddaiu, Giovannaⁿ, Topp, Cairistiona F.E.^o, Twardy, Stanislaw^p, Van
6 Middelkoop, Jantine^m, Wu, Lianhai^q and Bellocchi, Gianni^j

7 *Corresponding author

8 ^a IBERS, Aberystwyth University, 1st Floor, Stapledon Building, Plas Gogerddan, Aberystwyth, Ceredigion, UK, SY23 3EE.
9 Email: rpk@aber.ac.uk Tel: +441970 823160

10 ^b Green Technology, Natural Resources Institute Finland (Luke), Halolantie 31 A, 71750 Maaninka, Finland. Email:
11 perttu.virkajarvi@luke.fi

12 ^c Leibniz Universität Hannover, Institut für Gartenbauliche Produktionssysteme, Systemmodellierung Gemüsebau,
13 Herrenhäuser Straße 2, 30419 Hannover, Germany. Email: breitsameter@gem.uni-hannover.de

14 ^d Farming systems, territories and information technologies unit, Walloon agricultural research centre (CRA-W), 9 rue de
15 Liroux, B-5030 Gembloux, Belgium. Email: y.curnel@cra.wallonie.be

16 ^e ILVO, Plant Sciences Unit, Caritasstraat 39, 9090 Melle, Belgium. Email: tom.deswaef@ilvo.vlaanderen.be

17 ^f Swedish University of Agricultural Sciences (SLU), Department of Agricultural Research for Northern Sweden, SE-901 83
18 Umeå, Sweden. Email: Anne-Maj.Gustavsson@slu.se

19 ^g Norwegian Institute of Bioeconomy Research (NIBIO) Po. Box 115, NO-1431 Ås. Email: Tomas.Persson@nibio.no

20 ^h Arlon Campus Environnement, University of Liège, Avenue de Longwy 185, 6700 Arlon, Belgium. Email:
21 julien.minet@ulg.ac.be

22 ⁱ Institute of Landscape Systems Analysis, Leibniz Centre for Agricultural Landscape Research (ZALF), Eberswalder Straße 84,
23 15374 Müncheberg, Germany. Email: nendel@zalf.de

24 ^j UREP, INRA, 63000 Clermont-Ferrand, France. Email: gianni.bellocchi@clermont.inra.fr

25 ^k Potsdam Institute for Climate Impact Research, Telegraphenberg A31, 14473 Potsdam, Germany. Email: rolinski@pik-
26 potsdam.de

27 ^l Cranfield University, School of Energy, Environment, and Agri-food, College Road, Cranfield, Bedfordshire, MK43 0AL.

28 ^m Wageningen UR Livestock Research, P.O. Box 338, 6700 AH Wageningen, The Netherlands. Email:
29 jantine.vanmiddelkoop@wur.nl

30 ⁿ NRD, Desertification Research Centre; Dept. of Agriculture, University of Sassari, Viale Italia 39, 07100 - Sassari, Italy.
31 Email: gseddaiu@uniss.it

32 ^o SRUC, West Mains Road, Edinburgh, UK, EH9 3JG. Email: Kairsty.Topp@sruc.ac.uk

33 ^p Institute of Technology and Life Sciences at Falenty, Malopolska Research Centre in Krakow, 31-450 Krakow, ul. Ulanow
34 21B, Poland. Email: itepkrak@itep.edu.pl

35 ^q Rothamsted Research, North Wyke, Okehampton, UK, EX20 2SB. Lianhai.Wu@rothamsted.ac.uk

36

37

38

39

40 **Abstract**

41 Grassland-based ruminant production systems are integral to sustainable food production in Europe,
42 converting plant materials indigestible to humans into nutritious food, while providing a range of
43 environmental and cultural benefits. Climate change poses significant challenges for such systems,
44 their productivity and the wider benefits they supply. In this context, grassland models have an
45 important role in predicting and understanding the impacts of climate change on grassland systems,
46 and assessing the efficacy of potential adaptation and mitigation strategies. In order to identify the
47 key challenges for European grassland modelling under climate change, modellers and researchers
48 from across Europe were consulted via workshop and questionnaire. Participants identified fifteen
49 challenges and considered the current state of modelling and priorities for future research in
50 relation to each. A review of literature was undertaken to corroborate and enrich the information
51 provided during the horizon scanning activities. Challenges were in four categories relating to: 1) the
52 direct and indirect effects of climate change on the sward 2) climate change effects on grassland
53 systems outputs 3) mediation of climate change impacts by site, system and management and 4)
54 cross-cutting methodological issues. While research priorities differed between challenges, an
55 underlying theme was the need for accessible, shared inventories of models, approaches and data,
56 as a resource for stakeholders and to stimulate new research. Developing grassland models to
57 effectively support efforts to tackle climate change impacts, while increasing productivity and
58 enhancing ecosystem services, will require engagement with stakeholders and policy-makers, as well
59 as modellers and experimental researchers across many disciplines. The challenges and priorities
60 identified are intended to be a resource 1) for grassland modellers and experimental researchers, to
61 stimulate the development of new research directions and collaborative opportunities, and 2) for
62 policy-makers involved in shaping the research agenda for European grassland modelling under
63 climate change.

64

65 **Keywords**

66 Climate change; grasslands; horizon scanning; livestock production; models; research agenda

67

68

69

70

71 **1. Introduction**

72 The agricultural sector is facing unprecedented challenges as it attempts to maintain food security in
73 the context of climate and socio-economic change (Soussana, 2014; Thornton, 2010). The forecasted
74 increase of world population, dietary changes towards increasing meat consumption and the
75 demand for bioenergy suggest a global requirement for agricultural products by 2050 roughly twice
76 that of today (Foley et al., 2011). At the same time as increasing production, the livestock sector will
77 need to improve efficiency (Thornton, 2010) to avoid increasing the 26% of global land area
78 currently used for livestock production, and to reduce its estimated 15% share of total
79 anthropogenic greenhouse gas (GHG) emissions (Ripple et al., 2014). Havlik et al. (2014) suggest that
80 transitions from grass-based to more intensive livestock production systems may represent a cost-
81 effective approach to mitigating GHG emissions from livestock agriculture. However, while grass-
82 based ruminant production systems may be less efficient in terms of GHG emissions and land use
83 than more intensive systems, they provide a range of other benefits; European grasslands store an
84 estimated 5.5 Gt of carbon in the top 30 cm of their soils (Lugato et al., 2014). Covering around 30%
85 of agricultural land in Europe (Huyghe et al., 2014), grasslands also play an important role in the
86 maintenance of biodiversity and the sustenance of rural communities and cultures (Soussana and
87 Lemaire, 2014). Intensification or conversion of grasslands to crop production can lead to the
88 reduction or loss of such benefits (Dusseux et al., 2015). At the same time, ruminants valorise
89 marginal production areas, converting plant materials indigestible to humans into meat and dairy
90 products with high efficiency in terms of the consumption of human-edible food per unit of product
91 (Wheeler and Reynolds, 2013; Wilkinson, 2011). In Europe, around 25% of livestock protein intake
92 comes from grasslands (Leip et al., 2011). Despite these benefits, grasslands have declined in
93 Europe, with an estimated loss of seven million hectares between 1967 and 2007 (Huyghe et al.,
94 2014). Recent predictions suggest that this decline may continue in a climate change future (Leclère
95 et al., 2013). In this context, a better understanding is required of the impacts of climate change on
96 European grassland systems, the efficacy of adaptation strategies to increase their resilience and
97 productivity, and the pathways available to maintain and enhance the essential ecosystem services
98 they provide (Scollan et al., 2010; Smith et al., 2013).

99 In light of the challenges described, modelling can offer valuable support to farm and policy level
100 decision-makers, by providing tools to explore the performance of biophysical, management and
101 policy systems in the context of future climatic and socio-economic scenarios (Graux et al., 2013;
102 Kipling et al., 2014). A number of high-level strategic assessments of agricultural research priorities
103 (ATF, 2013; 2014; FACCE-JPI, 2012; Soussana, 2014) present a range of challenges to the agricultural

104 modelling community (Kipling et al., Accepted). The aim of this paper is to lay out in detail the
105 specific challenges and research priorities that grassland modelling must address, if it is to fulfil its
106 potential role in helping to tackle the global problems faced by the livestock production sector. The
107 focus of the paper is on European grasslands, and covers both permanent grasslands and leys
108 (grasslands established for less than five years). Three broad types of model applied to European
109 grasslands have previously been identified (Bellocchi et al., 2013); specialised grassland models, crop
110 models with grassland options, and vegetation models that can characterise a range of plant
111 communities including grasslands. This paper incorporates challenges relevant for all of these model
112 types, and explores links to other modelling disciplines and approaches.

113

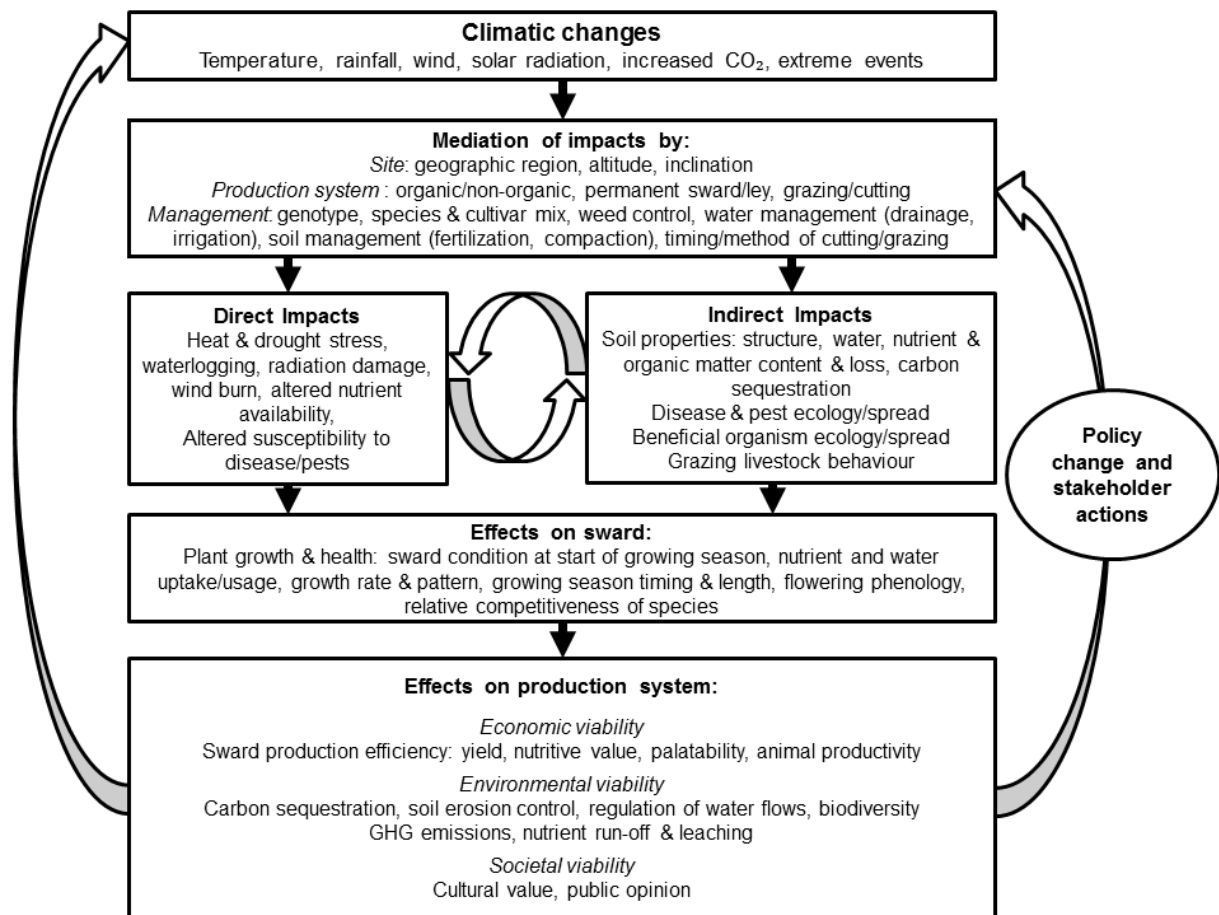
114 **2. Methods**

115 In order to understand the challenges and research priorities for grassland modelling, a ‘horizon
116 scanning’ approach based on that of Pretty et al. (2010) was used to gain the views of grassland
117 modellers and researchers from 18 institutes across 10 countries. The experts were drawn from, or
118 known to, partners contributing to a large European modelling network, the Agriculture, Food
119 Security and Climate Change Joint Programming Initiative (FACCE JPI) knowledge hub Modelling
120 European Agriculture with Climate Change for Food Security (MACSUR) (<http://www.macsur.eu>).
121 Views were gathered using a workshop and questionnaire and corroborated through the literature,
122 with the scope of discussions determined through a pre-workshop mapping process.

123 **2.1. Mapping Process**

124 Grassland models can cover a range of systems and processes, and a scoping exercise was necessary
125 to define the boundaries for discussions and questionnaire responses. Workshop facilitators and task
126 leaders involved in relevant activities within the MACSUR project created a single page diagram
127 intended to capture the components, processes and interactions associated with grassland
128 modelling. Participants were then asked to comment on and amend the map in an iterative process,
129 until a consensus was reached. The final map (Fig. 1) was used as a reference in workshop
130 discussions and distributed along with the questionnaire to guide responses.

131



132

133 **Figure 1:** Map of impacts of climate change on grassland systems, including feedbacks.

134 2.2. Workshop Approach

135 A workshop was held between the 17th and 19th of June 2015 at Wageningen University and
 136 Research Centre (The Netherlands). Workshop sessions were organised based on the ‘Futures
 137 Workshop’ approach (Jungk and Müllert, 1987; Valqui Vidal, 2005) as adapted for use in the EU FP7
 138 SOLID (Sustainable Organic and Low Input Dairying) project (<http://www.solidairy.eu>). Workshop
 139 participants were divided into small groups (5-6 people) and were invited to identify challenges to
 140 modelling in the subject areas covered by the workshop. Each participant wrote down as many
 141 challenges as they wished. Asking contributors to write down their suggestions ensured that all
 142 views were taken into account, reducing the problem of bias towards the opinions of the most vocal
 143 participants, which has been recognised in some focus group settings (Kitzinger, 1995). In discussion
 144 with their group, facilitators brought similar challenges together to remove duplication, and
 145 arranged them logically according to identified links between topics. Secondly, groups identified the
 146 ‘ideal world’ that would exist if each individual challenge were overcome. In the third step,
 147 participants were asked to discuss the current position and the potential for moving towards the
 148 ideal state for each challenge. Participants then identified practical research steps that could be

149 taken in each case. Finally, the small groups were brought together to exchange views and add
150 further comments and thoughts to the ‘maps’ created. The approach enabled a structured set of
151 challenges, research priorities and ideal world conditions to emerge from discussions of complex
152 topics encompassing many different disciplines and viewpoints.

153 2.3. Questionnaire approach and synthesis of outputs

154 In order that views could be gathered from experts who could not attend the workshop, a
155 questionnaire was designed using a similar structure to the workshop exercises and distributed to
156 contributors (Appendix 1). The questionnaire asked respondents to list challenges to modelling, ideal
157 states and the research steps required to move towards those ideals. Workshop outputs and
158 questionnaire responses were combined in a single spreadsheet, removing duplicated challenges
159 while retaining all distinct research steps identified. Information was shared with participants to
160 provide another opportunity for them to add to the challenges and research steps defined, based on
161 1) the development of their thoughts following initial participation, and 2) consideration of their
162 workshop and questionnaire responses in the context of existing literature. This round of revision
163 enabled descriptions of the current state of research to be enriched with reference to existing
164 review and research papers. The final list of challenges, ideal states and research steps were then
165 grouped into overarching themes.

166

167 3. Challenges and priorities for modelling

168 The workshop and questionnaire responses identified fifteen challenges. Twelve of these could be
169 categorized using the different aspects of grassland systems under climate change depicted in Fig. 1,
170 and three were cross-cutting challenges (Table 1). The first category of challenges relate to ‘direct
171 and indirect climate change effects on the sward’. Challenges one to three refer to biophysical
172 interactions which will require improved modelling in the context of climate change. These are
173 followed by three challenges (four to six) relating to modelling plant responses to climatic change,
174 while challenge seven considers the importance of widening the scope of modelling to take account
175 of pests and pathogens, the impact of which is likely to alter as the environment changes. The
176 category ‘Climate change effects on grassland system outputs’ (challenges eight to 10) focuses on
177 how environmental changes affect the economic and environmental outputs of grassland systems.
178 Challenges 11 to 12 in the category ‘Mediation of climate change impacts by site, system and
179 management’ cut across individual biophysical aspects, and are related to increasing capacity in
180 modelling different and changing systems, regions and management regimes. Finally, challenges 13

181 to 15 underpin the others, centring on making models that can adapt to stakeholder demands and
 182 overcoming technical and data-related challenges. These groups of challenges and priorities are
 183 described in the following section. The main lessons drawn from the challenges are then brought
 184 together (section 4).

185 **Table 1:** Challenges for grassland modelling identified by experts. Except for the methodological challenges,
 186 categories map onto the aspects of grassland systems depicted in Fig. 1. Challenges numbered as in the text.

Category	Challenge
Direct and indirect effects of climate change on the sward	1 Modelling multi-species swards
	2 Modelling soil variables/processes
	3 Modelling livestock and pasture interactions
	4 Modelling plant responses to environmental change
	5 Modelling overwintering
	6 Modelling the impact of extreme events
	7 Incorporating plant pests & pathogens into models
Climate change effects on grassland system outputs	8 Modelling the provision of ecosystem services
	9 Modelling nutrient cycles and GHG balances
	10 Modelling nutritional variables required to predict animal performance
Mediation of climate change impacts by site, system & management	11 Modelling different regions and production systems
	12 Modelling adaptation strategies
Cross-cutting methodological challenges	13 Making models 'fit-for-purpose'
	14 Linking different scales of modeling and data
	15 Providing data for models

187

188 *3.1. Direct and indirect effects of climate change on the sward*

189

190 1. Modelling multi-species swards

191 *The challenge:* Species-diverse swards may improve grassland resilience to changing climatic
 192 conditions (MacDougall et al., 2013). However, biodiversity, which has been linked to the provision
 193 of ecosystem services, may be affected by climate change, as relationships (both competitive and
 194 mutualistic) between species alter in novel and more variable conditions (Tylianakis et al., 2008;
 195 Vicca et al., 2006). Many grassland models were designed for application to single species swards, or

196 | to simple mixes such as clover and ryegrass (Lazzarotto et al., 2009). As a result, they are often
197 | limited in their capacity to characterise interactions in multi-species swards. These types of
198 | interaction may be complex, including above and below ground processes (Blomqvist et al., 2000;
199 | Dhamala et al., 2015) and transfers of nitrogen from legumes to other species (Nyfeler et al., 2011;
200 | Pirhofer-Walzl et al., 2011). There is growing recognition of the importance of understanding better
201 | the role of groups such as legumes in mixed swards, with a need for high protein forages to reduce
202 | reliance on expensive supplementary feeds and reduce nitrogen inputs (Lüscher et al., 2014; Suter et
203 | al., 2015).

204 | Some current process based models incorporate species mixtures to some extent (Ma et al., 2015)
205 | but further development is needed for uses that require characterisation beyond the definition of an
206 | average vegetation, for example in relation to the simulation of changes in sward composition. Snow
207 | et al. (2014) considered the ability of six grassland models to characterise multi-species swards,
208 | finding a diverse range of approaches to this challenge. They highlighted potential limitations in
209 | modelling more diverse swards, in the capacity of simpler approaches to adequately represent the
210 | impacts of changed conditions, and in the capacity to model novel species mixtures, such as swards
211 | including tree and shrub species. In the context of climate change, improving modelling capability in
212 | these respects is of particular importance, because of the expected changes in environmental
213 | conditions, increases in extreme events (challenge 6) and adaptation strategies incorporating
214 | increased sward diversity and agro-forestry (challenge 12).

215 | *Research priorities:* A full review of current modelling capability, data and knowledge relating to
216 | multi-species grasslands is required as a first step in defining the options for developing modelling
217 | capacity, including a theoretical framework for new multi-species models. Outputs and approaches
218 | from the vegetation modelling community can provide important insights with respect to
219 | interactions between species or functional types and their responses to climate change (Scheiter et
220 | al., 2013). An exploration of work on plant functional groups to identify the most important traits
221 | and processes (parameters) for modelling would ideally be a part of such a review. The most
222 | important types of sward for modellers to focus on could be investigated by reviewing information
223 | on the species mixtures that (based on current knowledge) are believed to perform best under
224 | climate change. Through the development of modular modelling approaches (challenge 13)
225 | connecting biodiversity modules to existing models offers one potential route to improve modelling
226 | capacity in relation to multi-species swards (challenge 8). Inventories of grassland models have been
227 | compiled as part of the activities of current research networks such as MACSUR (Bellocchi et al.,
228 | 2013) and comparisons of models such as that undertaken by Snow et al. (2014) provide the basis

229 for a more systematic synthesis of information about current models. Online repositories such as the
230 Agricultural Modelling Knowledge Hub (AgriMod) (<http://agrimod.basedev.co.uk>) can be used to
231 share such information, allowing model developers to update entries as their models are improved
232 over time.

233

234 2. Modelling soil variables/processes

235 *The challenge:* Many grassland models include fairly sophisticated ways of representing physical,
236 chemical and biological soil processes (Bellocchi et al., 2013). However, a range of complex
237 processes occur within the soil across many variables, including soil capillarity, leaching, evaporation,
238 effects of soil biota (such as earthworms), changes in the seed bank, soil microbial activity, impacts
239 of manuring and other fertilisation, and changes in soil organic matter. In the context of climate
240 change, experimental research and modelling has often focussed on the impacts of individual
241 variables affecting soil processes (soil warming, nitrogen deposition, water availability, CO₂
242 fertilization and fire) whilst it is known that interactions between such variables mean that their
243 combined effects are not easily predictable (Sierra et al., 2015). There are also complex interactions
244 between plants, mesofauna (Rossetti et al., 2015) and microbial populations and activity (Bagella et
245 al., 2014; Steinauer et al., 2015). Dunbabin et al. (2013) reviewed root architectural modelling and
246 identified the need for more data and conceptual models relating to soil biology, rhizosphere
247 chemistry, soil texture and mycorrhizas, as well as the need to consider root anatomy in models.

248 The development of SPACSYS (Wu et al., 2007) demonstrates how mechanistic plant (including root)
249 modelling can be applied at the field scale, while Perveen et al. (2014) describe the characterisation
250 in the SYMPHONY model of the impact of microbial diversity and the soil priming effect (the increase
251 in soil organic matter decomposition after fresh organic input) on soil-plant interactions. Linking root
252 modelling to soil models and engaging with plant modellers to drive real-world change (such as
253 improving plant genomes or predicting plant responses to change in the field) has been recognised
254 as a priority by the root modelling community (Dunbabin et al., 2013).

255 *Research priorities:* The preceding discussion indicates the need and scope for better communication
256 between grassland modellers, specialised soil and root modellers and experimental researchers, to
257 ensure that grassland models incorporate best practice in these disciplines, with as much detail as
258 needed to effectively fulfil the functions required of them (challenge 13). Contacts through networks
259 such as MACSUR, joint workshops, conference participation, and the development of infrastructure
260 for exchanging information could all support improved communication. Undertaking assessments of

261 the validity of the various functions and approaches used in modelling specific soil processes also
262 represents an important priority in reducing model uncertainty (Sierra et al., 2015). Improved
263 modelling of soil and hydrological processes is considered further in the context of modelling
264 nutrient cycles and GHG balances (challenge 9).

265

266 3. Modelling livestock and pasture interactions

267 *The challenge:* The impacts of livestock on grasslands, and the reciprocal impacts of grassland
268 management on livestock are multi-faceted and complex. In mixed swards, selective grazing by
269 animals and the spatial distribution of excreta can affect plant species composition and
270 characteristics, through direct influences on inter-specific competition, and indirectly through the
271 uneven distribution of nutrients (Liu et al., 2015; Xi et al., 2014). Grazing intensity is likely to affect
272 soil water retention, poaching, compaction (challenge 2), nutrient leaching and run-off, and GHG
273 emissions (challenge 9). Under conditions where the interaction between animal behaviour and the
274 environment have severe impacts on the sward, the effects on both grassland and livestock become
275 a function of management choices, as grazing pressure is reduced or animals are moved off the
276 pasture. In turn, sward composition, plant cover and condition directly affect feed availability and
277 digestibility (Hopkins and Wilkins, 2006), while external conditions, grazing behaviour and
278 management choices can all affect the disease and parasite risk from the grassland environment
279 (Fox et al., 2013; Smith et al., 2009). Models need to capture such relationships in order to identify
280 the best animal species, breeds and management regimes to maximise the efficiency of grassland-
281 based production under climate change in different environments. Snow et al. (2014) review the
282 various aspects of modelling livestock-pasture interactions, highlighting the challenges relating to
283 the trade-off between model usability and accuracy when attempting to model grazing interactions
284 at animal level, taking into account all the physical variables affecting forage intake. They conclude
285 that complex models are more important when grazing pressure is low (more extensive systems)
286 and in model uses where such detail is needed to model the subsequent digestion of the forage. The
287 importance of the challenges to improving modelling of livestock-pasture interactions is therefore
288 related to the purpose of the modelling effort (challenge 13) and the nature of the system (challenge
289 11).

290 *Research priorities:* Creating an inventory of the impacts of livestock on grassland (and the feedback
291 effects of grassland on livestock) for different livestock species and systems, and mapping this onto
292 the current capabilities of models, were seen by participants as important first steps to improve

293 modelling capacity. The biggest challenges are likely for models focussing on more extensive systems
294 with more diverse swards, because for these systems modelling is more complex, both on the animal
295 and the grassland side of the interaction. The described inventory can facilitate model comparisons,
296 the identification of gaps in knowledge and the testing of different approaches. As in other
297 challenges, improvements to allow both an accurate characterisation of livestock-pasture
298 interactions, and to understand how adaptation strategies might affect such interactions, will
299 require collaboration; in this case between grassland and livestock modellers (including animal
300 behaviour modellers) and between modellers and experimental researchers. Progress will be linked
301 to advances in modelling multi-species swards (challenge 1) and sward nutritive value (challenge 10).

302

303 4. Modelling plant responses to environmental change

304 *The challenge:* The quantification of plant responses to changing climate is a fundamental challenge
305 for crop grassland models. Climate change can affect grassland plants via changes in a range of
306 environmental conditions (Fig. 1) and plant responses are likely to vary with species and location
307 (Dumont et al., 2015). Plant responses to changes in climate include morphological and physiological
308 adaptation to stress and to raised CO₂ concentrations and changes in photosynthesis, biological
309 nitrogen fixation, and phenology; such responses involve changes in plant genes, proteins and
310 metabolites at different time-scales (Ahuja et al., 2010). White et al. (2012) highlighted variation in
311 methods and focus across experimental sites set up to study plant reactions to climate change, with
312 some impacts (temperature and water) studied more than others (such CO₂ and N addition) so that
313 results relating to individual impacts and interactions between impacts were hard to generalise. Only
314 a few experimental studies have investigated the combined effects of multiple environmental
315 stresses on grassland plants (Ahuja et al., 2010; Bertrand et al., 2008; Dieleman et al., 2012). Limits
316 to knowledge are therefore a constraint on model development in this research area. Current grass
317 and crop models characterise plant growth responses to a range of environmental impacts, including
318 changes in temperature, radiation, nitrogen and atmospheric CO₂ (Höglind et al., 2013; Wu et al.,
319 2007) including impacts on forage nutritive value (Ben Touhami et al., 2013; Jégo et al., 2013; Jing et
320 al., 2013; Thivierge et al., 2016). However, relatively few models incorporate all these aspects; some
321 processes (such as the impacts of CO₂ and variation in N) may be dealt with in a basic way, while some
322 interactions are not fully understood (Ramirez-Villegas et al., 2015). In relation to adaptive changes
323 in plant response over time, crop models have been used to explore the impacts of genetic
324 adaptation on yield under climate change conditions, and to define crop ideotypes for climate
325 change resilience (Rötter et al., 2015). However, Ramirez Villegas et al. (2015) highlighted

326 challenges, such as the need to couple genetic and crop models to produce outcomes suitable for
327 incorporation into breeding programmes, and the need to better quantify the robustness of model
328 outputs. In permanent swards with multiple species a range of factors including epigenetic and
329 plastic change and genetic change through natural selection and species sorting, shape grassland
330 responses to the environment. Inter-specific interactions may affect responses to climate change,
331 including changes in biomass production, sward composition and species diversity (Miranda-
332 Apodaca et al., 2015; Olsen et al., 2016). Improved modelling of these types of grassland depends on
333 the advancement of ecological knowledge, and progress in related topics including multi-species,
334 nutritive value and soil and water modelling (challenges 1, 2, 10).

335 *Research priorities:* Meta-experiments have been recommended to create international networks of
336 experimental sites which apply the same treatments and recording standards to investigate the
337 responses of swards to environmental change (Fraser et al., 2013; White et al., 2012). Over the long
338 term data from such programmes could facilitate more effective model improvement. Knowledge,
339 data and current model descriptions of the mechanisms underlying grassland plant responses should
340 be reviewed to assess capacity (which species are well characterised, which types of impact and
341 which interactions are incorporated and what are the limitations to the approaches used). This
342 should include consideration of how plant and field level responses are characterised in farm,
343 regional and global models, to evaluate effectiveness and areas for improvement. Ensemble model
344 exercises would be instructive in gaining an overview of current knowledge, including about the
345 climatic and regional boundaries within which grassland models work adequately (Soussana et al.,
346 2010). Drawing together such information would allow model development to be focused on the
347 most important relationships and interactions, in terms of their likely impact on grassland yield,
348 nutritive value and vulnerability to climate change. With respect to temporary grasslands, using
349 approaches used in crop modelling to explore resilient ideotypes for grassland species will be
350 important in better predicting the potential benefits of grass and legume breeding programmes in
351 climate change adaptation.

352

353 5. Modelling overwintering

354 *The challenge:* Modelling work with the aim of evaluating grassland performance often focuses on
355 the growing season. However, changes in permanent swards during the winter can, especially at
356 high latitudes and in mountainous regions, have important effects on subsequent productivity and
357 nutritional quality in spring and summer (Rapacz et al., 2014). Despite this, plant processes including,

358 hardening, de-hardening and re-hardening, vernalisation, winter respiration and allocation of
359 carbohydrates to reserve tissues (which can all affect the status of the sward during and after the
360 winter) are not sufficiently incorporated in most grassland models. As a result, the sensitivity of
361 grassland yield and nutritive quality to temperature variability, the frequency of extreme cold events
362 and snow cover depth, and management variables affecting winter performance (such as cutting
363 timing and frequency) cannot be satisfactorily assessed with current grassland models. A few
364 previous modelling attempts can serve as a basis for future efforts to improve the representation of
365 winter conditions in grassland models. These attempts include models, which simulate the cold
366 hardiness of winter wheat (Bergjord et al., 2008) and forage grass species (Thorsen and Höglind,
367 2010) as expressed by the temperature at which 50% of plants in a population die (i.e. the LT50
368 value). Changes to the LT50 value can be caused by hardening, de-hardening and re-hardening
369 processes during the winter season, which are a function of the prevailing temperature in the upper
370 soil layer surrounding the crown of the plant, and a cultivar-specific maximum hardiness parameter.
371 Snow cover models have also been linked to the STICS model for continuous multi-seasonal
372 simulations of annual spring crops in eastern Canada (Jégo et al., 2014). Recently, a full-year model
373 (BASGRA), for timothy grass was developed by combining a growing season model with cold-
374 hardening and soil physical models for the winter season (Höglind et al., Accepted).

375 *Research priorities:* An important next step for model development in this field will be to test the
376 winter-related functions of grassland models against data from experiments simulating projected
377 future winter conditions. Further model development in this field will depend on the availability of
378 experimental data on cold sensitivity and the state of the sward (such as tiller density and leaf, stem
379 and reserve weight during the growing season and over winter). As well as the collection of new
380 data, the systematic organization of existing datasets on these variables according to temperature,
381 precipitation and photoperiod gradients would be beneficial to the development and applicability of
382 winter modules across geographic regions and climatic conditions.

383

384 6. Modelling the impact of extreme events

385 *The challenge:* The impacts of extreme events on grassland productivity are of increasing concern in
386 relation to food security (Long and Ort, 2010) and the continuing supply of services from grassland
387 systems (Bloor and Bardgett, 2012). While models are improving in terms of their ability to predict
388 the impact of changes in average climate conditions on grassland yields, modelling the impact of
389 extreme events such as droughts, heatwaves, flooding and frost exposure, remains a challenge. A

390 unique definition of an extreme event is also difficult to formulate. Beyond the statistical occurrence
391 of an event exceeding a low or a high percentile threshold, an extreme weather event may be
392 defined as one that has a high impact on society and biophysical systems. Thus, it is a hard-to-predict
393 phenomenon far beyond normal expectations (Peterson et al., 2012). Different types of extreme
394 events often occur together, so that different plant stress factors (e.g. high temperature, low water
395 availability or flooding and waterlogging, evaporative demand and high light intensities) may affect
396 vegetation simultaneously and in different combinations across geographical areas. This generates
397 complexity in climate forcing / plant response relationships across a wide range of temporal and
398 spatial scales. The poor description of this complexity in current grassland models can lead to
399 inaccuracies in simulated processes (Soussana et al., 2010). These limitations become especially
400 apparent when the capacity of grassland plants to acclimate to harsh conditions is substantially
401 exceeded. For example, temperatures that are abnormally low or high often result in lower plant
402 productivity at all subsequent temperatures (Zaka et al., Accepted). In climate change impact studies
403 using grassland models, responses to extreme temperatures and prolonged water deficits are still
404 not sufficiently considered (Reyer et al., 2013; Ruppert et al., 2015). They are also scarce in model
405 calibration and validation datasets due to their low frequency in weather data time series (Ben
406 Touhami and Bellocchi, 2015). The mechanistic relationships between plant processes and the
407 impact of extreme events on these processes have only been fragmentarily documented, and the
408 extent to which plants may be able to respond to extreme weather events remains an open field of
409 research (Reyer et al., 2013). The many interactions between vegetation, soil and the atmosphere,
410 and the role of management practices make our ability to simulate grassland systems limited.
411 Predictions of the impact of extreme events therefore require accurate information about
412 management, animal behaviour and the prior condition of the sward, in addition to data on weather
413 conditions and methods for characterising the interactions between these variables. Few
414 experimental data relate to extreme conditions, with much information collected when long-term
415 monitoring captures the impacts of extreme events by chance (Thibault and Brown, 2008).

416 *Research priorities:* To improve modelling of the impacts of extreme events, a review of data and
417 gaps in knowledge in relation to the types of event expected to affect grasslands under climate
418 change is required, including an appraisal of current definitions of extreme events and the
419 thresholds which produce them. An inventory of the capabilities of existing grassland models in
420 relation to extreme events would enable limitations in current approaches to be identified, and
421 options for improvement developed. These could include the development of extreme events
422 functions (affecting transpiration, photosynthesis, tillering, resource allocation, etc.) that could be
423 linked to existing grassland models. Such functions can draw on knowledge from studies about

424 processes of dehydration and recovery of plant communities and functional types (Zwicke et al.,
425 2013) and the explicit representation of hydraulic processes (Tardieu et al., 2015) while also
426 addressing interactions with water and nitrogen cycling (Calanca et al., 2016). Data from ongoing
427 monitoring programmes will have an important role in model validation as new extreme events
428 occur. Grassland data relating to previous extreme events can also be examined to better
429 understand resilience. Current projects, such as MODEXTREME (<http://modextreme.org/>) and
430 MERINOVA (<https://merinova.vito.be/Pages/home.aspx>) offer collaborative arenas for making
431 progress in overcoming this challenge. The synthesis and sharing of outcomes from these projects in
432 the wider modelling community will be important in the future development of modelling capacity.

433

434 7. Incorporating plant pests and pathogens into models

435 *The challenge:* Pathogens and pests can affect crop and grassland yield in a range of ways (Gregory
436 et al., 2009). Climate change is expected to have complex impacts on crops and their interactions
437 with pathogens and pests, including increased plant vulnerability resulting from their genetic
438 responses to the effects of environmental change, changes in pest and pathogen fecundity and
439 growth rate, and changes in assemblages of pest antagonist species (Gregory et al., 2009; Rapacz et
440 al., 2014; Zulka and Götzl, 2015). These relationships are complex. Although interactions between
441 plants and pathogens in mixed species swards are not fully understood, there is evidence that
442 pathogens can play an important role in maintaining sward diversity and even in maintaining higher
443 productivity in diverse swards, with swards made up of few species more vulnerable to pests and
444 pathogens (Bever et al., 2015).

445 In general, grassland models do not incorporate the impacts of pests and pathogens currently
446 affecting European grasslands, nor the changes in pathogen spread expected as a consequence of
447 climate change. At present the characterisation of pathogens and pests in the modelling of leys is
448 fairly limited, for example assuming constraints based on the 'disease class' of different crops in crop
449 rotation models (Annetts and Audsley, 2002). Looking beyond insect and microbial pests and
450 pathogens, grazing by other species, such as waterfowl, can also cause significant problems for
451 grassland productivity (Merkens et al., 2012), and to the authors' knowledge, this has yet to be
452 addressed in grassland modelling.

453 *Research priorities:* Gregory et al. (2009) highlight the need for modelling the impacts of pests and
454 pathogens under climate change that takes into account complex interactions of these species with
455 other biotic and abiotic variables. This should go beyond current coupling of climate change and

456 weather-based disease forecasting, or the prediction of future pest and pathogen distributions
457 based on information about their ecological niches and climate mapping.

458 Further developing process-based modelling approaches is important to better understand the
459 impact of pathogens and pests under climate change conditions. In an example of this kind of
460 approach, Whish et al. (2015) combined two process-based models – a pathogen population model
461 (DYSIM) and the APSIM crop model – to investigate the impact of a wheat rust on yield. Such
462 mechanistic approaches may be used to provide the insights required to model more complex multi-
463 species interactions with pathogens. Assessing the impacts of adaptation measures, for example in
464 the form of resilient cultivars, changes in crop rotations or the conservation and development of
465 plant diversity in grasslands will also require improved knowledge of pest-pathogen interactions. A
466 further priority will be to model how plot-level interactions are mediated by landscape
467 characteristics; for example, the impacts of biodiverse semi-natural habitats which are known to
468 promote antagonist species of pests (Zulka and Götzl, 2015), linking to the idea of resilient Climate
469 Smart Landscapes (Scherr et al., 2012).

470 The collation of existing knowledge about key pests and pathogens of grasslands across different
471 regions, including information about their ecology (such as their likely response to climate change
472 and control by antagonist species) along with an assessment of models developed across disciplines
473 to investigate them, would be a first step to improving modelling capacity. Such an inventory could
474 be used as a basis to review the options for modelling the future effects of these pathogens under
475 climate change, in mono-cultures and in multi-species swards.

476

477 *3.2. Climate change effects on grassland systems outputs*

478

479 8. Modelling the provision of ecosystem services

480 *The challenge:* At present, many agricultural grassland models focus on productivity, without taking
481 into account the value of ecosystem services provided by grasslands (Kipling et al., Accepted). A
482 number of authors have identified a range of beneficial roles played by grassland systems (Hönigová
483 et al., 2012; Zhao et al., 2003) including: soil erosion control and rainfall regulation (critical in the
484 context of increased occurrence of extreme events under climate change; challenge 6), soil carbon
485 accumulation and nutrient cycling (challenge 9), air quality purification, biodiversity maintenance

486 and the sustaining of cultural diversity. In relation to each of these services, models need to be able
487 to characterise the impacts of climate change and associated changes in management strategies.

488 A range of modelling approaches is currently used to evaluate the impact of farm- and policy-level
489 decisions on biodiversity, and to incorporate biodiversity into multi-objective models at the regional
490 scale (Kipling et al., Accepted). There is also potential for, and some examples of, agricultural models
491 being used in conjunction with ecological models to explore interactions between production,
492 management choices and biodiversity (Tixier et al., 2013) while modelling tools are being developed
493 to evaluate grassland ecosystem services more generally (Campion et al., 2014). The need for more
494 research on carbon sequestration (challenge 9), water regulation and conservation of soils
495 (challenge 2) across EU climate regions has also been recognised (Soussana et al., 2004). Advances in
496 modelling these relationships rely on developments in experimental research to understand more
497 fully the mechanisms underlying the provision of ecosystem services and their relationship to
498 production (Pilgrim et al., 2010).

499 Given that ecological and social resilience to extreme events are intertwined (Adger, 2000) and that
500 diversity and modularity are important components of social resilience (Carpenter et al., 2012) the
501 role of grasslands in maintaining cultural diversity is no less important than the 'physical' services
502 discussed in the context of climate change. In this respect, developing the capacity to model
503 traditional extensive systems that have received less attention in the past (challenge 11) and
504 participatory engagement with stakeholders to develop relevant models and explore adaptation
505 alternatives, are important priorities (challenge 13).

506 *Research priorities:* Participants suggested that a first step towards the better characterisation in
507 grassland models of ecosystem services and the impacts of climate change upon them would be to
508 identify modelling capacity with respect to each pairing of ecosystem service and climate change
509 impact across different European regions. This process could draw on published work and reports on
510 ecosystem services, such as Hönigová et al. (2012), and climate change impacts, such as Iglesias et al.
511 (2012), and on model inventories currently available in the literature. This exercise should be
512 inclusive of ecology, vegetation, hydrology and soil models, to reveal not only gaps in capacity, but
513 also areas in which models from these different disciplines could be used together to provide
514 assessments of grassland systems encompassing the evaluation of non-commodified services.

515

516 9. Modelling nutrient cycles and GHG balances

517 *The challenge:* Modelling of GHG emissions from ruminant production systems has received much
518 attention, but challenges still remain in the characterisation of anaerobic slurry digestion and CH₄
519 leakage, NH₃ and N₂O emissions from manure, and the interaction of nitrogen with soil and weather
520 in relation to NO₃ leaching (Kipling et al., Accepted). Focusing on grasslands, understanding and
521 modelling soil processes is central to estimating nutrient flows (challenge 2).

522 -Reviewing models of carbon release arising from soil organic matter (SOM) decomposition, Sierra et
523 al. (2015) identified the need for more data on and better characterisation of SOM decomposition
524 processes at high temperature and extremes of moisture, and for a critical assessment of the range
525 of functions used to represent such processes in different models. Recent modelling by Perveen et
526 al. (2014) (see also challenge 2) incorporated the characterisation of the soil priming effect and
527 microbial diversity into the SYMPHONY model, and used it to examine impacts on soil and plant
528 interactions and carbon and nitrogen dynamics under climate change.

529 Studying combined impacts of environmental change on nutrient cycling, rather than the impact of
530 individual changes in isolation, is an important challenge to be met (Sierra et al., 2015). Recent
531 research has found that plant diversity may play a more important role than temperature in
532 determining the communities of microbes involved in carbon, nitrogen and phosphorous cycles
533 (Steinauer et al., 2015), and that the expected increase in soil carbon emissions arising from higher
534 temperatures may be mediated by consumption of fungi by soil invertebrates (Crowther et al.,
535 2015). These findings highlight the importance of considering biotic and abiotic processes together.
536 Increasing the capacity to model such interactions will therefore require collaboration between
537 modelling communities and with experimental researchers.

538 *Research Priorities:* Participants suggested that tests on the impacts of manure management on
539 emissions (for example, the method and timing of applications and manure type) were required to
540 support improved grassland modelling in this area, with more data on nitrogen fluxes and pools also
541 important. The development of models characterising closed nitrogen cycles and incorporating the
542 history of nitrogen in plants and the soil, was considered another priority for improving modelling
543 capacity. Overall, improving model equations relating to N₂O and CH₄ emissions, as well as improving
544 the definition of carbon pools, and work to relate N₂O emissions to the efficiency of nitrogen uptake
545 by plants in models, are important areas for development, with the aim of tackling some of the
546 complexity described in this section. These steps can help to reduce model uncertainty and increase
547 the capacity to model nutrient cycles and emissions under different climate change scenarios.

548

549 10. Modelling nutritional variables required to predict animal performance

550 *The challenge:* Modelling sward nutritional value (see also challenge 1) is of particular importance
551 for understanding the interactions between grasslands and livestock nutrition. Changes in nutritional
552 value will alter the need for other feeds and supplements and affect productivity and the quality of
553 final products. Impacts may also arise through altered intake by livestock caused by changes in
554 grazing behaviour (challenge 3). The nutritional value of ruminant feed includes a range of variables:
555 nitrogen fraction (total nitrogen, nitrogen solubility, nitrogen degradability, acid detergent insoluble
556 nitrogen); potentially fermentable fraction (water soluble carbohydrates, pectins, starch and cell
557 walls); non-fermentable fraction (volatile fatty acids, lactate, lipids) (AFRC, 1998). Climate change is
558 expected to affect the nutritive value of grassland swards through nutritional changes in individual
559 species, and changes in species composition, with impacts varying according to conditions (for
560 example mountain versus Mediterranean grasslands) and species type (Dumont et al., 2015). Where
561 grasslands are cut for silage, hay or in 'cut-and-carry' systems, rather than grazed directly by
562 livestock, nutritive value will also be affected by cutting time, and by subsequent treatment and
563 storage; climate change is expected to alter the optimal timing and number of silage cuts (in terms
564 of yield and nutritive value) per year in northern Europe (Höglind et al., 2013). Given this complexity,
565 the detail with which models characterise nutritive value must be tailored to reflect the aims of
566 individual modelling exercises (challenge 13). The modelling of changes in grassland yields (Graux et
567 al., 2013; Vital et al., 2013) is well developed. However, the characterisation of nutritive value in
568 grassland models has been in general limited to species-specific responses to conditions, for
569 example in timothy (Duru et al., 2010; Jégo et al., 2013) rather than changes in value in multi-species
570 swards (Kipling et al., Accepted).

571 *Research priorities:* Grassland and livestock modellers and animal nutritionists need to work
572 together to identify the most important nutritional parameters for incorporation into grassland
573 models in relation to different applications. This should include gaining an overview of the extent to
574 which current models are capable of characterising these parameters. Harmonising how nutritive
575 value is reported and calculated for modelling, and in model outputs, will also require cooperation,
576 with the aim of allowing models to be applied, compared and evaluated across Europe. These
577 collaborative developments can facilitate the creation of more models able to provide the
578 nutritional data required to support accurate predictions of animal performance under climate
579 change.

580

581 3.3. Mediation of climate change impacts by site, system and management

582

583 11. Modelling different regions and production systems

584 *The challenge:* Models are often developed to answer questions relating to specific systems within a
585 particular region. Llewellyn et al. (2007) found that stakeholders are most interested in local
586 information, and that presenting such information can aid understanding and uptake of modelled
587 solutions. As a result, models may not perform well when applied to other conditions. For example,
588 the focus of previous modelling has often been on intensive and non-organic systems, such as that
589 reported by Jing *et al.* (2012) and Jégo et al. (2013). In part, this may reflect the complexities of
590 modelling heterogeneous extensive swards likely to contain multiple species (challenge 1). There are
591 also gaps in the modelling of region-specific systems. For example, grassland models designed for
592 temperate systems mainly characterise perennial species, while Mediterranean grasslands are
593 dominated by annuals. In addition, perennial species in these systems undergo a period of summer
594 dormancy due to harsh conditions in the summer months. Although some models, such as STICS
595 (Ruget et al., 2009) consider summer dormancy in perennial species, relatively few models have
596 focussed on these types of grassland, despite the expected negative impact of climate change on
597 Mediterranean regions of Europe (Iglesias et al., 2012). In this case, the systems in question differ
598 between regions, but differences may also cut across regions.

599 *Research priorities:* In order to realise the ideal of having models able to predict climate change
600 impacts and the effectiveness of adaptation and mitigation strategies across systems and regions,
601 undertaking a systematic assessment of current capacity was considered important. This could be
602 achieved by using and further developing model inventories such as those created as part of the
603 MACSUR project (Bellocchi et al., 2013), in order to match models to the systems and regions they
604 were designed for, or could potentially be suitable for. Assessments of the potential for widening
605 model applicability can draw on the findings of investigations that have used generic approaches to
606 model biophysical processes across a variety of regions (Yuan et al., 2014). Recent work comparing
607 models from different regions, such as carried out within the FP7 project MultiSward
608 (http://www.multisward.eu/multisward_eng/) the MACSUR project (Sándor et al., 2015; 2016) and
609 the Agricultural Model Inter-comparison and Improvement Programme (AgMIP)
610 (<http://www.agmip.org>) can provide further evidence about the applicability of models to different
611 environments and systems. This baseline information could inform new modelling research and data
612 collection in order to fill identified gaps in capacity, and to ensure that climate change impacts are

613 effectively modelled across regions and systems. The applicability of models to other systems and
614 regions will depend on the characteristics of the focus system/region and of the model itself, but
615 also on the level of detail required to achieve specific aims (challenge 13).

616

617 12. Modelling adaptation strategies

618 *The challenge:* Modelling adaptation strategies requires both that the designs of models allow
619 changes in biophysical and/or economic variables to drive, and be driven by, management choices
620 over successive model cycles, and that reactions to changing circumstances realistically characterise
621 the behaviour of decision makers. The first part of this challenge therefore relates to the
622 development of capacity to model the physical impacts of grassland management such as, cutting
623 and grazing and interactions with re-growth and flowering, fertilization and interactions with pest
624 and disease susceptibility, changes in soil organic matter, and changes in the system being used, for
625 example, from mono-culture to mixed pasture or from permanent to temporary grassland.
626 Adaptation also includes plant breeding strategies (see challenge 4); models can be used to
627 investigate the traits or trait combinations of benefit for species under climate change in different
628 contexts. However, so far models have rarely been applied to grassland species (Van Oijen and
629 Höglind, 2015), and progress will require more data on the genetics of different plant traits, as well
630 as new model methodologies.

631 Models will need to characterise how different management strategies interact with other variables
632 and with outputs in terms of yield and quality; for example, the effect of a wet harvest season on
633 herbage and silage nutritional value and on associated costs, such as the need to buy supplementary
634 feeds. In this context, linking to other types of modelling will be important, for example to
635 characterise the livestock health and environmental risks associated with manure application given
636 expected climate-related changes in pathogen spread (Venglovsky et al., 2009). Recent models such
637 as PaturaMata have been specifically developed in order to design management strategies for farms
638 under climate change (Dusseux et al., 2015) and many current grassland models can be asked to
639 respond to specific changes. Some process based farm scale models, such as the Integrated Farm
640 Systems Model (Rotz et al., 2014) and some grassland models (Vuichard et al., 2007) are able to
641 explore the impact of different management strategies (such as changes in cutting regimes) under
642 climate change (Thivierge et al., 2016) but further development is required to improve the scope of
643 adaptation options covered, and the characterisation of interactions between different strategies
644 (Del Prado et al., 2013). Such development should take into account the need to explore the

645 potential of more 'explorative' adaptation strategies (Martin et al., 2013) such as the introduction of
646 silvo-pasture (Broom et al., 2013).

647 Adaptation includes not just changes of management, but also changes of system. At regional level,
648 economic land use models have been applied to forecast changes in agricultural land use as a result
649 of climatic and socio-economic changes, based on profit thresholds for different land uses (Audsley
650 et al., 2015). As farmers' choices about the adoption of adaptation strategies are known to be
651 affected by both economic and non-economic considerations (for example, their perception of
652 climate change risks) (Llewellyn, 2007; Lyle, 2015) the second part of this challenge (to more
653 accurately characterise the uptake of adaptation strategies) is also complex.

654 *Research priorities:* To develop the capacity of models to characterise the impacts of adaptation
655 strategies will initially require the collation of resources detailing available strategies for different
656 systems and regions, such as provided by Iglesias et al. (2012) and Iglesias and Garrote (2015),
657 including current knowledge related to their efficacy and the mechanisms via which they work.
658 Assessments can then be made of the availability and limitations of modelling in relation to different
659 strategies and their potential interactions with other management and policy decisions. Options for
660 incorporating current understanding of stakeholder decision-making into bio-physical models need
661 to be explored, in order to ensure that models better characterise the likely uptake of adaptation
662 strategies. One approach would be to use the identified adaptation strategies to develop context-
663 dependent adaptation scenarios, fitted to the expectations and knowledge of relevant stakeholders.
664 Finally, management modules (as well as the characterisation of biophysical relationships) will need
665 to be validated for climate change conditions.

666

667 3.4. Cross-cutting methodological challenges

668

669 13. Making models 'fit-for-purpose'

670 *The challenge:* The different contexts in which grassland models are used require those models to
671 have very different characteristics, in terms of complexity (including the types and resolution of data
672 they require; challenge 15), the scales of inputs required (challenge 14) and outputs delivered, and
673 the level of capacity to model management changes and stakeholder choices (see also challenge 12).
674 Mechanistic models have great value for understanding more about complex processes and
675 interactions, while at larger scales and for more practical applications simpler mechanistic and
676 empirical models, informed by this deeper understanding, can be effective predictive tools.

677 Therefore, the apparent trade-off between model usability and accuracy can be seen instead as an
678 iterative development process (Kipling et al., Accepted). In this context, the type of model applied to
679 a particular problem should reflect the nature of the problem and the needs of the stakeholders
680 concerned (Ramirez-Villegas et al., 2015). This can be achieved through the iterative involvement of
681 relevant stakeholders in model development and evaluation (Bellocchi et al., 2015). To achieve the
682 best outcomes, stakeholders should also be able to easily choose between available modelling tools,
683 requiring them to be shared and packaged to allow comparison of their usefulness in different
684 contexts (Voinov and Bousquet, 2010). Modelling platforms which support the development of
685 interchangeable sub-models, can produce modular modelling tools that are easily adapted for
686 specific and emerging uses (Holzworth et al., 2015). In crop and grassland modelling, the Biophysical
687 Models Applications (BioMA) framework (<http://bioma.jrc.ec.europa.eu>) is a good example of a
688 software platform that supports modular model development and evaluation.

689 *Research priorities:* A key first step to developing more adaptable models is to gain an overview of
690 their current capabilities in relation to different potential uses. Creating a checklist style inventory
691 which clearly compares model applicability in relation to specific tasks would both highlight scales
692 and types of modelling that are missing, and help stakeholders and policy-makers to select the most
693 appropriate modelling tools to support their activities. Model inventories within projects such as
694 MACSUR (Bellocchi et al., 2013) form the basis for the development of such a resource, while online
695 hubs such as Agrimod provide the potential to share this information with wider scientific and
696 stakeholder communities. A checklist inventory could be a starting point for reviewing the options
697 for developing further flexibility and accessibility. While modular modelling and open access
698 modelling can be valuable, the challenges to collaborative working need to be recognised in a
699 competitive scientific environment. In this context, a resource presenting existing and developing
700 tools in a format accessible to stakeholders may create more favourable conditions for mutual
701 learning between modellers, while maintaining the valuable diversity required to tackle climate
702 change related issues which can vary by region and system (challenge 11).

703

704 14. Linking different scales of modelling and data

705 *The challenge:* Grassland simulations can be defined at different spatial scales ranging from plot to
706 region. Input data are often supplied, and output data may be produced, at different scales than that
707 at which the analysis is performed, thus requiring the application of down- or up-scaling techniques
708 (Höglind et al., 2013). The level of detail of input and output data varies with the model (and often

709 with the country) and thus the required level of upscaling / downscaling. The spatial extent and
710 resolution of data is therefore a critical issue which must be accorded special attention (Zhao et al.,
711 2015) considering that changing spatial resolution by aggregation or disaggregation of data (e.g.
712 using field-scale impact models with input data at scales other than that for which they were
713 developed) bears the risk of missing the relevant scale of a process or phenomenon. Specifically,
714 climate models produce large scale output data while micro-climatic changes can be important for
715 grassland modelling. Extrapolations of local soil properties to larger regions can also help assess the
716 requirement for soil input in regional estimations (Persson et al., 2015). Insufficient automation of
717 composition and execution, and scalability of approaches can be one of the reasons for the absence
718 of comprehensive, computer-aided, and spatiotemporal assessments. This is true especially in local
719 contexts where automated procedures become essential to link downscaled climate scenarios to
720 biophysical outputs and socio-economic impacts (Walz et al., 2014).

721 *Research priorities:* The systematic evaluation of the software and techniques available for down-
722 scaling of data is required in order to understand the limitations and strengths of the different
723 approaches, and to gain insight into the scale dependence of grassland models (Zhao and Liu, 2014).
724 Better access for modellers to down-scaling techniques is also important, alongside evidence on
725 their performance. In addition, systematic tests of model sensitivity to changes in data resolution,
726 including in relation to climate data, are important in order to establish where scaling techniques, or
727 the provision of data at a different resolution, would be most beneficial. Eza et al. (2015) describe
728 the application of a modelling platform for climate change vulnerability studies (and their
729 incorporation into management and planning), where grassland simulation capabilities are at the
730 core of integrated and automated procedures (including down- and up-scaling approaches) usually
731 employed in isolation.

732

733 15. Providing data for models

734 *The challenge:* Models rely on experimental data for their development, evaluation and application
735 to different problems. Data issues vary for different areas of grassland modelling. They can be
736 categorised as 1) The need for data from new experimental work 2) Quality and completeness of
737 available data, 3) Data accessibility, and 4) Variation in data measurement and recording:

738 1) Datasets which include information about previous management (for example, the age
739 of the grassland, previous fertilisation, cutting or grazing) are often lacking, for example
740 in relation to data on soil carbon and carbon sequestration. In general there have been

741 fewer studies investigating interactions between variables, for example in studies of soil
742 processes (challenge 2) with a focus on single variables more usual. Modelling can
743 increase understanding of complex systems and the interactions within them (Van
744 Paassen et al., 2007). In this way models can highlight priorities for future experimental
745 research. Developing the relationship between modellers and experimental researchers
746 can therefore drive well-focussed experimental research and data collection (Kipling et
747 al., 2014).

748 2) The detailed information required for some aspects of grassland modelling can be
749 obtained from experimental sites set up for long term data collection, such as
750 micrometeorological flux measurement sites (Baldocchi et al., 2001). However, data
751 from other sources need better evaluation in terms of the methods used, their
752 compatibility with specific models, and the level of detail they include. Through the
753 MACSUR knowledge hub, Kersebaum et al. (2015) developed a quantitative classification
754 framework to evaluate the quality and consistency of existing agricultural datasets for
755 use in crop models. This framework is likely to be applicable for the identification of data
756 for grassland models, especially for models used to characterise both grassland and
757 cropping systems (Bellocchi et al., 2013). New approaches to data collection include the
758 use of remote sensing (Courault et al., 2010; Verrelst et al., 2015) and the development
759 of virtual weather stations that combine a range of data sources to improve rainfall
760 estimates (Racca et al., 2011). These advances can improve data accuracy and provide
761 new data-sources of potential value for grassland modelling.

762 3) Open access data platforms such as FLUXNET (Baldocchi et al., 2001) provide examples
763 of how standardised collecting, processing and delivery of data can be developed, and
764 that data shared. In other areas, online resources to share meta-data have been created,
765 for example for soil data at European and global levels (Kipling et al., 2015) and sites
766 specifically focused on sharing information about models and data such as Agrimod
767 provide important resources for grassland modellers.

768 4) Differences between nations and research groups in the way that variables are
769 measured and recorded can cause problems, for example, differences in the definitions
770 of forage nutrient values (challenge 10) can hinder the use of data for modelling.
771 Differences in terminology and approach have been recognised as barriers to inter-
772 disciplinary collaboration (Siedlok and Hibbert, 2014), and overcoming them requires
773 enhanced communication and understanding between researchers across Europe. The

774 implementation of standardised collection, processing and delivery of data is particularly
775 important when undertaking model inter-comparison studies.

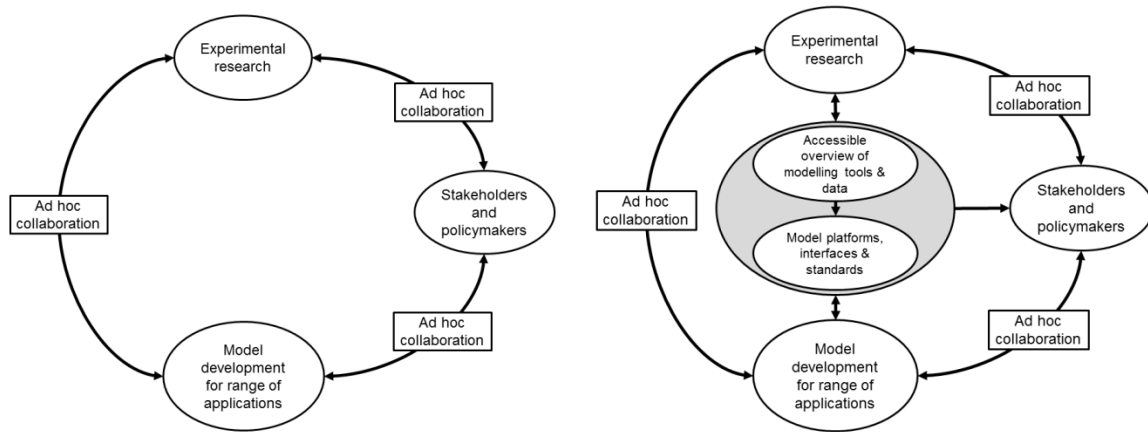
776 *Research priorities:* Improved communication between modelling groups and experimental
777 researchers is vital to ensure that shared meta-data on available datasets allows their identification
778 and evaluation for use by grassland modellers. This will require modellers to effectively
779 communicate the data types and standards that they require, developing and sharing protocols for
780 data evaluation such as those described in this section. The need for such developments is common
781 to a range of agricultural modelling disciplines, and inter-disciplinary collaboration is therefore vital
782 in this area to prevent duplication of effort. Networks such as MACSUR, AgMIP and the Global
783 Research Alliance (<http://globalresearchalliance.org>) are essential in providing arenas in which
784 modellers can collaborate to create and enhance these community resources. The development of
785 networks of experimental sites and coordinated experiments across nations to investigate climate
786 change impacts on grasslands (White et al., 2012) would also support model development, by
787 providing high quality, comparable data.

788

789 **4. Synthesis**

790 The fifteen challenges for grassland modelling identified here (Table 1) cover all aspects of
791 modelling. Although many of the challenges have been discussed in previous reviews, such as Bryant
792 and Snow (2008), Snow et al. (2014) and Holzworth et al. (2015), to the authors' knowledge this has
793 been the first attempt to comprehensively assess the challenges and priorities for European
794 grassland modelling in the context of climate change, using a collaborative horizon scanning
795 approach. In identifying the research priorities associated with each modelling challenge,
796 participants repeatedly highlighted the need for a clear and comprehensive collation and sharing of
797 information on current grassland modelling tools and methodological approaches. Across the
798 challenges considered, the benefit of such resources to drive both the development of modelling on
799 specific topics, and the development of more adaptable, accessible modelling platforms and
800 approaches was highlighted. These priorities suggest that, despite the development of a range of
801 research networks and collaborative groupings relating to agricultural modelling, a high degree of
802 compartmentalisation still exists between researchers in different research groups, institutes and
803 nations. As well as spurring and focussing the development of new experimental and modelling
804 research, rich, shared inventories of models and data are also important for stakeholders and policy-
805 makers seeking the most relevant modelling tools to meet their needs (challenge 13). Access to

806 effective modelling tools is a vital element of supporting stakeholders in making effective decisions
 807 (Voinov and Bousquet, 2010). The current state of grassland modelling can be illustrated by ad hoc
 808 interactions between modellers, experimental researchers and stakeholders (Fig 2, left panel).
 809 Addressing the modelling priorities identified in this exercise would move the community towards
 810 greater coherence, with shared model and data inventories driving research and collaboration, and
 811 supporting stakeholder choices (Fig 2, right panel).



812

813 **Figure 2:** Modelling, experimental research, and stakeholder interactions without community resources (left)
 814 and with community resources (right).

815 Across the agricultural research community, the need for joined up approaches to tackling the issues
 816 of climate change have long been appreciated (Soussana et al., 2012) and current network initiatives
 817 are starting to move agricultural modellers towards the realisation of a more joined-up, focussed
 818 modelling community, as some of the resources developed in MACSUR, GRA and AgMIP (Antle et al.,
 819 2015; Bellocchi et al., 2013; Kersebaum et al., 2015; Yeluripati et al., 2015) demonstrate. However,
 820 long term support and governance will be required if these efforts are to be successfully extended
 821 (Kipling et al., Accepted) given the barriers to scientific collaboration, especially across disciplines
 822 (Siedlok and Hibbert, 2014). While initiatives such as MACSUR have been shown to have a positive
 823 impact on levels of collaborative engagement, there also appears to be more work to do to engage
 824 with researchers beyond a well-connected core (Saetnan and Kipling, Accepted) and to provide the
 825 more comprehensive and accessible resources for grassland modellers and stakeholders described
 826 here.

827 In relation to the more specific challenges for European grassland modelling, the need to learn from
 828 advances in other fields was a noticeable component of many research priorities, for example: the
 829 incorporation of understanding and approaches from soil and root modelling (challenge 2 and 9),
 830 from livestock modelling (challenges 3 and 10), from plant and ecosystem modelling (challenge 1, 4,

831 5, 8) and from those involved in research and modelling of stakeholder decision-making (challenge
832 12). Across the challenges relating to individual climate change impacts, the reliance of grassland
833 models on the availability of suitable data (challenge 15) for further development was also clear.
834 Finally, meeting the methodological challenges (13 – 15) will require technical dialogue between
835 modelling disciplines which might successfully adopt the same methods despite widely differing
836 subject matter. Better sharing and comparisons of models presented in accessible inventories, the
837 subsequently improved visibility of opportunities for collaboration (Fig 2) and networking between
838 disciplines, will be required to make these types of link in an effective way.

839 A horizon scanning approach has allowed the collation of views of grassland modellers and
840 researchers from across Europe, while subsequent consideration of the literature validated opinions
841 expressed in the workshop session and via questionnaire. It is hoped that the presentation of these
842 findings will help grassland modellers to identify new directions and collaborative opportunities in
843 their research, and guide policy makers involved in shaping the research agenda for European
844 grassland modelling under climate change.

845

846 **5. Conclusions**

847 The horizon scanning exercise presented in this paper identified 15 challenges to European grassland
848 modelling in the context of climate change (Table 1), considered the current state of modelling in
849 relation to each challenge, and presented pathways to improving model capacity. The responses of
850 participants to this exercise highlighted the need for the creation of shared resources within the
851 grassland modelling community, in order to 1) allow stakeholders to identify and select modelling
852 tools to suit their needs, and 2) drive experimental and modelling research by focussing attention on
853 gaps in knowledge and opportunities for collaboration (including engagement with stakeholders
854 during model development). The creation of such resources will require long-term support and
855 governance in order to overcome the barriers to such cooperative endeavours in a competitive
856 scientific environment. However, the complex, multi-faceted nature of climate change makes such
857 developments essential.

858 **Acknowledgements**

859 This paper was supported by the FACCE-JPI knowledge hub MACSUR with national funding from
860 BBSRC and Scottish Government (UK), EL&I (The Netherlands), INRA (France), MIPAAF (Italy), MMM
861 (Finland), RCN (Norway), SPW (Belgium), The National Centre for Research and Development

862 (Poland), FORMAS (Sweden), JÜLICH and BLE (Germany). The authors would like to thank Dr Panu
863 Korhonen (Luke) and two anonymous reviewers for their contributions to the revision of this paper.

864 **References**

- 865 Adger WN. Social and ecological resilience: are they related? *Progress in Human Geography* 2000;
866 24: 347-364. doi: 10.1191/030913200701540465
- 867 AFRC. Technical Committee on Responses to Nutrients, report no. 11. CABI, Wallingford, 1998.
- 868 Ahuja I, de Vos RCH, Bones AM, Hall RD. Plant molecular stress responses face climate change.
869 *Trends in Plant Science* 2010; 15: 664-674. doi: 10.1016/j.tplants.2010.08.002
- 870 Annetts JE, Audsley E. Multiple objective linear programming for environmental farm planning. *The*
871 *Journal of the Operational Research Society* 2002; 53: 933-943. doi:
872 10.1057/palgrave.jors.2601404
- 873 Antle JM, Valdivia RO, Boote KJ, Janssen S, Jones JW, Porter CH, et al. AgMIP's transdisciplinary
874 agricultural systems approach to regional integrated assessment of climate impacts,
875 vulnerability, and adaptation. In: Rosenzweig C, Hills D, editors. *Handbook of Climate Change*
876 *and Agroecosystems*. Imperial College Press, London, 2015.
- 877 ATF. Research and innovation for a sustainable livestock sector in Europe: Suggested priorities for
878 support under Horizon 2020 to enhance innovation and sustainability in the animal
879 production sector of Europe's food supply chains. Animal Task Force white paper, 2013.
- 880 ATF. Research and innovation for a competitive and sustainable animal production sector in Europe:
881 Recommended priorities for support under Horizon 2020 in the 2016/2017 programme. 1st
882 Addendum to the Animal Task Force white paper, 2014.
- 883 Audsley E, Trnka M, Sabaté S, Maspons J, Sanchez A, Sandars D, et al. Interactively modelling land
884 profitability to estimate European agricultural and forest land use under future scenarios of
885 climate, socio-economics and adaptation. *Climatic Change* 2015; 128: 215-227. doi:
886 10.1007/s10584-014-1164-6
- 887 Bagella S, Filigheddu R, Caria MC, Girlanda M, Roggero PP. Contrasting land uses in Mediterranean
888 agro-silvo-pastoral systems generated patchy diversity patterns of vascular plants and
889 below-ground microorganisms. *Comptes Rendus Biologies* 2014; 337: 717-724. doi:
890 10.1016/j.crv.2014.09.005
- 891 Baldocchi D, Falge E, Gu L, Olson R, Hollinger D, Running S, et al. FLUXNET: A new tool to study the
892 temporal and spatial variability of ecosystem-scale carbon dioxide, water vapor, and energy
893 flux densities. *Bulletin of the American Meteorological Society* 2001; 82: 2415-2434. doi:
894 10.1175/1520-0477(2001)082<2415:fantts>2.3.co;2
- 895 Bellocchi G, Ma S, Köchy M, Braunmiller K. Identified grassland-livestock production systems and
896 related models. *FACCE MACSUR Reports* 2013; 2: D-L2.1.1. doi:
897 Bellocchi G, Rivington M, Matthews K, Acutis M. Deliberative processes for comprehensive
898 evaluation of agroecological models. A review. *Agronomy for Sustainable Development*
899 2015; 35: 589-605. doi: 10.1007/s13593-014-0271-0
- 900 Ben Touhami H, Bellocchi G. Bayesian calibration of the Pasture Simulation model (PaSim) to
901 simulate European grasslands under water stress. *Ecological Informatics* 2015; 30: 356-364.
902 doi: 10.1016/j.ecoinf.2015.09.009
- 903 Ben Touhami H, Lardy R, Barra V, Bellocchi G. Screening parameters in the Pasture Simulation model
904 using the Morris method. *Ecological Modelling* 2013; 266: 42-57. doi:
905 10.1016/j.ecolmodel.2013.07.005
- 906 Bergjord AK, Bonesmo H, Skjelvåg AO. Modelling the course of frost tolerance in winter wheat: I.
907 Model development. *European Journal of Agronomy* 2008; 28: 321-330. doi:
908 10.1016/j.eja.2007.10.002

909 Bertrand A, Tremblay GF, Pelletier S, Castonguay Y, Belanger G. Yield and nutritive value of timothy
910 as affected by temperature, photoperiod and time of harvest. *Grass and Forage Science*
911 2008; 63: 421-432. doi: 10.1111/j.1365-2494.2008.00649.x

912 Bever JD, Mangan SA, Alexander HM. Maintenance of plant species diversity by pathogens. *Annual*
913 *Review of Ecology, Evolution, and Systematics* 2015; 46: 305-325. doi: 10.1146/annurev-
914 *ecolsys-112414-054306*

915 Blomqvist MM, Olff H, Blaauw MB, Bongers T, Van Der Putten WH. Interactions between above- and
916 belowground biota: importance for small-scale vegetation mosaics in a grassland ecosystem.
917 *Oikos* 2000; 90: 582-598. doi: 10.1034/j.1600-0706.2000.900316.x

918 Bloor JMG, Bardgett RD. Stability of above-ground and below-ground processes to extreme drought
919 in model grassland ecosystems: Interactions with plant species diversity and soil nitrogen
920 availability. *Perspectives in Plant Ecology, Evolution and Systematics* 2012; 14: 193-204. doi:
921 10.1016/j.ppees.2011.12.001

922 Broom DM, Galindo FA, Murgueitio E. Sustainable, efficient livestock production with high
923 biodiversity and good welfare for animals. *Proceedings of the Royal Society of London B:*
924 *Biological Sciences* 2013; 280. doi: 10.1098/rspb.2013.2025

925 Bryant JR, Snow VO. Modelling pastoral farm agro-ecosystems: A review. *New Zealand Journal of*
926 *Agricultural Research* 2008; 51: 349-363. doi: 10.1080/00288230809510466

927 Calanca P, Deléglise C, Martin R, Carrère P, Mosimann E. Testing the ability of a simple grassland
928 model to simulate the seasonal effects of drought on herbage growth. *Field Crops Research*
929 2016; 187: 12-23. doi: 10.1016/j.fcr.2015.12.008

930 Champion M, Ninane M, Hautier L, Dufrêne M, Stilmant D. BIOECOSYS: towards the development of a
931 decision support tool to evaluate grassland ecosystem services. In: Hopkins A, Collins R,
932 Fraser M, King V, Lloyd D, Moorby J, et al., editors. *EGF at 50: The future of European*
933 *grasslands. Proceedings of the 25th General Meeting of the European Grassland Federation.*
934 19. Prysogol Aberystwyth, Aberystwyth, 2014, pp. 376-378.

935 Carpenter SR, Arrow KJ, Barrett S, Biggs R, Brock WA, Crépin A-S, et al. General resilience to cope
936 with extreme events. *Sustainability* 2012; 4: 3248-3259. doi: 10.3390/su4123248

937 Courault D, Hadria R, Ruget F, Oliosio A, Duchemin B, Hagolle O, et al. Combined use of FORMOSAT-2
938 images with a crop model for biomass and water monitoring of permanent grassland in
939 Mediterranean region. *Hydrology and Earth System Sciences Discussions* 2010; 14: 1731-
940 1744. doi: 10.5194/hess-14-1731-2010

941 Crowther TW, Thomas SM, Maynard DS, Baldrian P, Covey K, Frey SD, et al. Biotic interactions
942 mediate soil microbial feedbacks to climate change. *Proceedings of the National Academy of*
943 *Sciences* 2015; 112: 7033-7038. doi: 10.1073/pnas.1502956112

944 Del Prado A, Crosson P, Olesen JE, Rotz CA. Whole-farm models to quantify greenhouse gas
945 emissions and their potential use for linking climate change mitigation and adaptation in
946 temperate grassland ruminant-based farming systems. *animal* 2013; 7: 373-385. doi:
947 doi:10.1017/S1751731113000748

948 Dhamala NR, Sjøgaard K, Eriksen J. Competitive forbs in high-producing temporary grasslands with
949 perennial ryegrass and red clover can increase plant diversity and herbage yield. *Grassland*
950 *Science in Europe, Volume 20. Wageningen Academic Publishers, Wageningen, 2015, pp.*
951 *209-211.*

952 Dieleman WIJ, Vicca S, Dijkstra FA, Hagedorn F, Hovenden MJ, Larsen KS, et al. Simple additive
953 effects are rare: a quantitative review of plant biomass and soil process responses to
954 combined manipulations of CO₂ and temperature. *Global Change Biology* 2012; 18: 2681-
955 2693. doi: 10.1111/j.1365-2486.2012.02745.x

956 Dumont B, Andueza D, Niderkorn V, Lüscher A, Porqueddu C, Picon-Cochard C. A meta-analysis of
957 climate change effects on forage quality in grasslands: specificities of mountain and
958 Mediterranean areas. *Grass and Forage Science* 2015; 70: 239-254. doi: 10.1111/gfs.12169

959 Dunbabin VM, Postma JA, Schnepf A, Pagès L, Javaux M, Wu L, et al. Modelling root–soil interactions
 960 using three–dimensional models of root growth, architecture and function. *Plant and Soil*
 961 2013; 372: 93-124. doi: 10.1007/s11104-013-1769-y
 962 Duru M, Cruz P, Martin G, Theau JP, Charron M-H, Desange M, et al. Herb'sim: un modèle pour
 963 raisonner la production et l'utilisation de l'herbe. *Fourrages* 2010; 201: 37-46. doi:
 964 Dusseux P, Zhao Y, Cordier M-O, Corpetti T, Delaby L, Gascuel-Odoux C, et al. PaturMata, a model to
 965 manage grassland under climate change. *Agronomy for Sustainable Development* 2015; 35:
 966 1087-1093. doi: 10.1007/s13593-015-0295-0
 967 Eza U, Shtiliyanova A, Borrás D, Bellocchi G, Carrère P, Martin R. An open platform to assess
 968 vulnerabilities to climate change: An application to agricultural systems. *Ecological*
 969 *Informatics* 2015; 30: 389-396. doi: 10.1016/j.ecoinf.2015.10.009
 970 FACCE-JPI. Strategic Research Agenda, 2012.
 971 Foley J, Ramankutty N, Brauman K, Cassidy E, Gerber J, Johnston M, et al. Solutions for a cultivated
 972 planet. *Nature* 2011; 478: 337-342. doi: 10.1038/nature10452
 973 Fox NJ, Marion G, Davidson RS, White PCL, Hutchings MR. Modelling parasite transmission in a
 974 grazing system: The importance of host behaviour and immunity. *PLoS ONE* 2013; 8: e77996.
 975 doi: 10.1371/journal.pone.0077996
 976 Fraser LH, Henry HAL, Carlyle CN, White SR, Beierkuhnlein C, Cahill JF, et al. Coordinated distributed
 977 experiments: an emerging tool for testing global hypotheses in ecology and environmental
 978 science. *Frontiers in Ecology and the Environment* 2013; 11: 147-155. doi: 10.1890/110279
 979 Graux A-I, Bellocchi G, Lardy R, Soussana J-F. Ensemble modelling of climate change risks and
 980 opportunities for managed grasslands in France. *Agricultural and Forest Meteorology* 2013;
 981 170: 114-131. doi: 10.1016/j.agrformet.2012.06.010
 982 Gregory PJ, Johnson SN, Newton AC, Ingram JSI. Integrating pests and pathogens into the climate
 983 change/food security debate. *Journal of Experimental Botany* 2009; 60: 2827-2838. doi:
 984 10.1093/jxb/erp080
 985 Havlík P, Valin H, Herrero M, Obersteiner M, Schmid E, Rufino MC, et al. Climate change mitigation
 986 through livestock system transitions. *Proceedings of the National Academy of Sciences* 2014;
 987 111: 3709-3714. doi: 10.1073/pnas.1308044111
 988 Höglind M, Thorsen SM, Semenov MA. Assessing uncertainties in impact of climate change on grass
 989 production in Northern Europe using ensembles of global climate models. *Agricultural and*
 990 *Forest Meteorology* 2013; 170: 103-113. doi: 10.1016/j.agrformet.2012.02.010
 991 Höglind M, Van Oijen M, Cameron C, Persson T. Process-based simulation of growth and
 992 overwintering of grassland using the BASGRA model. *Ecological Modelling* Accepted. doi:
 993 Holzworth DP, Snow V, Janssen S, Athanasiadis IN, Donatelli M, Hoogenboom G, et al. Agricultural
 994 production systems modelling and software: Current status and future prospects.
 995 *Environmental Modelling & Software* 2015; 72: 276-286. doi: 10.1016/j.envsoft.2014.12.013
 996 Hönigová I, Vačkář D, Lorencová E, Melichar J, Götzl M, Sondregger G, et al. Survey on grassland
 997 ecosystem services. Report to the EEA - European Topic Centre on Biological Diversity.
 998 Nature Conservancy Agency of the Czech Republic, Prague, 2012.
 999 Hopkins A, Wilkins RJ. Temperate grassland: key developments in the last century and future
 1000 perspectives. *The Journal of Agricultural Science* 2006; 144: 503-523. doi:
 1001 10.1017/S0021859606006496
 1002 Huyghe C, De Vlieghe A, van Gils B, Peeters A. Grasslands and herbivore production in Europe and
 1003 effects of common policies. Versailles: Editions Quae, 2014.
 1004 Iglesias A, Garrote L. Adaptation strategies for agricultural water management under climate change
 1005 in Europe. *Agricultural Water Management* 2015; 155: 113-124. doi:
 1006 10.1016/j.agwat.2015.03.014
 1007 Iglesias A, Quiroga S, Moneo M, Garrote L. From climate change impacts to the development of
 1008 adaptation strategies: Challenges for agriculture in Europe. *Climatic Change* 2012; 112: 143-
 1009 168. doi: 10.1007/s10584-011-0344-x

1010 Jégo G, Belanger G, Tremblay GF, Jing Q, Baron VS. Calibration and performance evaluation of the
1011 STICS crop model for simulating timothy growth and nutritive value. *Field Crops Research*
1012 2013; 151: 65-77. doi: 10.1016/j.fcr.2013.07.003

1013 Jégo G, Chantigny M, Pattey E, Belanger G, Rochette P, Vanasse A, et al. Improved snow-cover model
1014 for multi-annual simulations with the STICS crop model under cold, humid continental
1015 climates. *Agricultural and Forest Meteorology* 2014; 195-196: 38-51. doi:
1016 10.1016/j.agrformet.2014.05.002

1017 Jing Q, Belanger G, Baron V, Bonesmo H, Virkajarvi P. Simulating the Nutritive Value of Timothy
1018 Summer Regrowth. *Agronomy Journal* 2013; 105: 563-572. doi: 10.2134/agronj2012.0331

1019 Jing Q, Bélanger G, Baron V, Bonesmo H, Virkajärvi P, Young D. Regrowth simulation of the perennial
1020 grass timothy. *Ecological Modelling* 2012; 232: 64-77. doi: 10.1016/j.ecolmodel.2012.02.016

1021 Jungk R, Müllert N. *Future Workshops: How to Create Desirable Futures*. London: Institute for Social
1022 Inventions, 1987.

1023 Kersebaum KC, Boote KJ, Jorgenson JS, Nendel C, Bindi M, Frühauf C, et al. Analysis and classification
1024 of data sets for calibration and validation of agro-ecosystem models. *Environmental*
1025 *Modelling & Software* 2015; 72: 402-417. doi: 10.1016/j.envsoft.2015.05.009

1026 Kipling RP, Bannink A, Bellocchi G, Dalgaard T, Fox NJ, Hutchings NJ, et al. Modelling European
1027 ruminant production systems: facing the challenges of climate change. *Agricultural Systems*
1028 Accepted. doi:

1029 Kipling RP, Saetnan E, Scollan N, Bartley D, Bellocchi G, Hutchings NJ, et al. Modelling livestock and
1030 grassland systems under climate change. In: Hopkins A, Collins R, Fraser M, King V, Lloyd D,
1031 Moorby J, et al., editors. *EGF at 50: The future of European grasslands*. Proceedings of the
1032 25th General Meeting of the European Grassland Federation. Prifysgol Aberystwyth,
1033 Aberystwyth, 2014, pp. 97-99.

1034 Kipling RP, Topp K, Don A. The availability of carbon sequestration data in Europe. *FACCE MACSUR*
1035 *Reports* 2015; 4: D-L1.4.2. doi:

1036 Kitzinger J. Qualitative research. Introducing focus groups. *British Medical Journal* 1995; 311: 299-
1037 302. doi:

1038 Lazzarotto P, Calanca P, Fuhrer J. Dynamics of grass-clover mixtures-An analysis of the response to
1039 management with the PROductive GRASSland Simulator (PROGRASS). *Ecological Modelling*
1040 2009; 220: 703-724. doi: 10.1016/j.ecolmodel.2008.11.023

1041 Leclère D, Jayet P-A, de Noblet-Ducoudré N. Farm-level autonomous adaptation of European
1042 agricultural supply to climate change. *Ecological Economics* 2013; 87: 1-14. doi:
1043 10.1016/j.ecolecon.2012.11.010

1044 Leip A, Achermann B, Billen G, Bleeker A, Bouwman AF, de Vries W, et al. Integrating nitrogen fluxes
1045 at the European scale. In: Sutton MA, Howard CM, Erisman JW, Billen G, Bleeker A, Grennfelt
1046 P, et al., editors. *The European Nitrogen Assessment: Sources, Effects and Policy*
1047 *Perspectives*. Cambridge University Press, Cambridge, 2011, pp. 345-376.

1048 Liu J, Feng C, Wang D, Wang L, Wilsey BJ, Zhong Z. Impacts of grazing by different large herbivores in
1049 grassland depend on plant species diversity. *Journal of Applied Ecology* 2015; 52: 1053-1062.
1050 doi: 10.1111/1365-2664.12456

1051 Llewellyn RS. Information quality and effectiveness for more rapid adoption decisions by farmers.
1052 *Field Crops Research* 2007; 104: 148-156. doi: 10.1016/j.fcr.2007.03.022

1053 Long SP, Ort DR. More than taking the heat: crops and global change. *Current Opinion in Plant*
1054 *Biology* 2010; 13: 240-247. doi: 10.1016/j.pbi.2010.04.008

1055 Lugato E, Panagos P, Bampa F, Jones A, Montanarella L. A new baseline of organic carbon stock in
1056 European agricultural soils using a modelling approach. *Global Change Biology* 2014; 20:
1057 313-326. doi: 10.1111/gcb.12292

1058 Lüscher A, Mueller-Harvey I, Soussana JF, Rees RM, Peyraud JL. Potential of legume-based
1059 grassland–livestock systems in Europe: a review. *Grass and Forage Science* 2014; 69: 206-
1060 228. doi: 10.1111/gfs.12124

1061 Lyle G. Understanding the nested, multi-scale, spatial and hierarchical nature of future climate
1062 change adaptation decision making in agricultural regions: A narrative literature review.
1063 *Journal of Rural Studies* 2015; 37: 38-49. doi: 10.1016/j.jrurstud.2014.10.004
1064 Ma S, Lardy R, Graux, A.-I. BT, H., Klumpp, K., Martin, R., Bellocchi, G. Regional-scale analysis of
1065 carbon and water cycles on managed grassland systems ☆. 2015. doi:
1066 10.1016/j.envsoft.2015.03.007
1067 MacDougall AS, McCann KS, Gellner G, Turkington R. Diversity loss with persistent human
1068 disturbance increases vulnerability to ecosystem collapse. *Nature* 2013; 494: 86-89. doi:
1069 10.1038/nature11869
1070 Martin G, Martin-Clouaire R, Duru M. Farming system design to feed the changing world. A review.
1071 *Agronomy for Sustainable Development* 2013; 33: 131-149. doi: 10.1007/s13593-011-0075-4
1072 Merkens M, Bradbeer DR, Bishop CA. Landscape and field characteristics affecting winter waterfowl
1073 grazing damage to agricultural perennial forage crops on the lower Fraser River delta, BC,
1074 Canada. *Crop Protection* 2012; 37: 51-58. doi: 10.1016/j.cropro.2012.02.014
1075 Miranda-Apodaca J, Pérez-López U, Lacuesta M, Mena-Petite A, Muñoz-Rueda A. The type of
1076 competition modulates the ecophysiological response of grassland species to elevated CO₂
1077 and drought. *Plant Biology* 2015; 17: 298-310. doi: 10.1111/plb.12249
1078 Nyfeler D, Huguenin-Elie O, Suter M, Frossard E, Lüscher A. Grass–legume mixtures can yield more
1079 nitrogen than legume pure stands due to mutual stimulation of nitrogen uptake from
1080 symbiotic and non-symbiotic sources. *Agriculture, Ecosystems & Environment* 2011; 140:
1081 155-163. doi: 10.1016/j.agee.2010.11.022
1082 Olsen SL, Töpper JP, Skarpaas O, Vandvik V, Klanderud K. From facilitation to competition:
1083 temperature-driven shift in dominant plant interactions affects population dynamics in
1084 seminatural grasslands. *Global Change Biology* 2016; 22: 1915-1926. doi: 10.1111/gcb.13241
1085 Persson T, Kværnø S, Höglind M. Impact of soil type extrapolation on timothy grass yield under
1086 baseline and future climate conditions in southeastern Norway. *Climate Research* 2015; 65:
1087 71-86. doi: 10.3354/cr01303
1088 Perveen N, Barot S, Alvarez G, Klumpp K, Martin R, Rapaport A, et al. Priming effect and microbial
1089 diversity in ecosystem functioning and response to global change: a modeling approach
1090 using the SYMPHONY model. *Global Change Biology* 2014; 20: 1174-1190. doi:
1091 10.1111/gcb.12493
1092 Peterson TC, Stott PA, Herring S. Explaining extreme events of 2011 from a climate perspective.
1093 *Bulletin of the American Meteorological Society* 2012; 93: 1041-1067. doi: 10.1175/bams-d-
1094 12-00021.1
1095 Pilgrim ES, Macleod JA, Blackwell MSA, Bol R, Hogan DV, Chadwick DR, et al. Interactions among
1096 agricultural production and other ecosystem services delivered from European temperate
1097 grasslands. *Advances in Agronomy* 2010; 109: 117-154. doi: 10.1016/S0065-2113(10)09004-
1098 8
1099 Pirhofer-Walzl K, Rasmussen J, Høgh-Jensen H, Eriksen J, Sørensen K, Rasmussen J. Nitrogen transfer
1100 from forage legumes to nine neighbouring plants in a multi-species grassland. *Plant and Soil*
1101 2011; 350: 71-84. doi: 10.1007/s11104-011-0882-z
1102 Pretty J, Sutherland WJ, Ashby J, Auburn J, Baulcombe D, Bell M, et al. The top 100 questions of
1103 importance to the future of global agriculture. *International Journal of Agricultural*
1104 *Sustainability* 2010; 8: 219-236. doi: 10.3763/ijas.2010.0534
1105 Racca P, Kleinhenz B, Zeuner T, Keil B, Tschöpe B, Jung J. Decision support systems in agriculture:
1106 administration of meteorological data, use of geographic information systems (GIS) and
1107 validation methods in crop protection warning service. In: Jao C, editor. *Efficient Decision*
1108 *Support Systems - Practice and Challenges From Current to Future*. InTech, Rijeka, 2011, pp.
1109 331-354.
1110 Ramirez-Villegas J, Watson J, Challinor AJ. Identifying traits for genotypic adaptation using crop
1111 models. *Journal of Experimental Botany* 2015. doi: 10.1093/jxb/erv014

1112 Rapacz M, Ergon Å, Höglind M, Jørgensen M, Jurczyk B, Østrem L, et al. Overwintering of herbaceous
1113 plants in a changing climate. Still more questions than answers. *Plant Science* 2014; 225: 34-
1114 44. doi: 10.1016/j.plantsci.2014.05.009

1115 Reyer CPO, Leuzinger S, Rammig A, Wolf A, Bartholomeus RP, Bonfante A, et al. A plant's perspective
1116 of extremes: terrestrial plant responses to changing climatic variability. *Global Change*
1117 *Biology* 2013; 19: 75-89. doi: 10.1111/gcb.12023

1118 Ripple WJ, Smith P, Haberl H, Montzka SA, McAlpine C, Boucher DH. Ruminants, climate change and
1119 climate policy. *Nature Clim. Change* 2014; 4: 2-5. doi: 10.1038/nclimate2081

1120 Rossetti I, Bagella S, Cappai C, Caria MC, Lai R, Roggero PP, et al. Isolated cork oak trees affect soil
1121 properties and biodiversity in a Mediterranean wooded grassland. *Agriculture, Ecosystems &*
1122 *Environment* 2015; 202: 203-216. doi: 10.1016/j.agee.2015.01.008

1123 Rötter RP, Tao F, Höhn JG, Palosuo T. Use of crop simulation modelling to aid ideotype design of
1124 future cereal cultivars. *Journal of Experimental Botany* 2015. doi: 10.1093/jxb/erv098

1125 Ruget F, Satger S, Volaire F, Lelièvre F. Modeling tiller density, growth, and yield of Mediterranean
1126 perennial grasslands with STICS. *Crop Science* 2009; 49: 2379-2385. doi:
1127 10.2135/cropsci2009.06.0323

1128 Ruppert JC, Harmony K, Henkin Z, Snyman HA, Sternberg M, Willms W, et al. Quantifying drylands'
1129 drought resistance and recovery: the importance of drought intensity, dominant life history
1130 and grazing regime. *Global Change Biology* 2015; 21: 1258-1270. doi: 10.1111/gcb.12777

1131 Saetnan E, Kipling RP. Evaluating a knowledge hub: are we building a better connected community?
1132 *Scientometrics* Accepted. doi:

1133 Sándor R, Acutis M, Barcza Z, Ben Touhami H, Doro L, Hidy D, et al. Sensitivity and uncertainty
1134 analysis of grassland models in Europe and Israel. *FACCE MACSUR Reports 2* 2015; 5: SP5-55.
1135 doi:

1136 Sándor R, Barcza Z, Hidy D, Lellei-Kovács E, Ma S, Bellocchi G. Modelling of grassland fluxes in
1137 Europe: Evaluation of two biogeochemical models. *Agriculture, Ecosystems & Environment*
1138 2016; 215: 1-19. doi: 10.1016/j.agee.2015.09.001

1139 Scheiter S, Langan L, Higgins SI. Next-generation dynamic global vegetation models: learning from
1140 community ecology. *New Phytologist* 2013; 198: 957-969. doi: 10.1111/nph.12210

1141 Scherr S, Shames S, Friedman R. From climate-smart agriculture to climate-smart landscapes.
1142 *Agriculture & Food Security* 2012; 1: 1-15. doi: 10.1186/2048-7010-1-12

1143 Scollan ND, Greenwood PL, Newbold CJ, Ruiz DRY, Shingfield KJ, Wallace RJ, et al. Future research
1144 priorities for animal production in a changing world. *Animal Production Science* 2010; 51: 1-
1145 5. doi: 10.1071/AN10051

1146 Siedlok F, Hibbert P. The organization of interdisciplinary research: Modes, drivers and barriers.
1147 *International Journal of Management Reviews* 2014; 16: 194-210. doi: 10.1111/ijmr.12016

1148 Sierra CA, Trumbore SE, Davidson EA, Vicca S, Janssens I. Sensitivity of decomposition rates of soil
1149 organic matter with respect to simultaneous changes in temperature and moisture. *Journal*
1150 *of Advances in Modeling Earth Systems* 2015; 7: 335-356. doi: 10.1002/2014ms000358

1151 Smith J, Sones K, Grace D, MacMillan S, Tarawali S, Herrero M. Beyond milk, meat, and eggs: Role of
1152 livestock in food and nutrition security. *Animal Frontiers* 2013; 3: 6-13. doi: 10.2527/af.2013-
1153 0002

1154 Smith LA, Marion G, Swain DL, White PCL, Hutchings MR. The effect of grazing management on
1155 livestock exposure to parasites via the faecal–oral route. *Preventive Veterinary Medicine*
1156 2009; 91: 95-106. doi: 10.1016/j.prevetmed.2009.05.026

1157 Snow VO, Rotz CA, Moore AD, Martin-Clouaire R, Johnson IR, Hutchings NJ, et al. The challenges –
1158 and some solutions – to process-based modelling of grazed agricultural systems.
1159 *Environmental Modelling & Software* 2014; 62: 420-436. doi: 10.1016/j.envsoft.2014.03.009

1160 Soussana J-F. Research priorities for sustainable agri-food systems and life cycle assessment. *Journal*
1161 *of Cleaner Production* 2014; 73: 19-23. doi: 10.1016/j.jclepro.2014.02.061

1162 Soussana J-F, Fereres E, Long SP, Mohren FGMJ, Pandya-Lorch R, Peltonen-Sainio P, et al. A
1163 European science plan to sustainably increase food security under climate change. *Global*
1164 *Change Biology* 2012; 18: 3269-3271. doi: 10.1111/j.1365-2486.2012.02746.x

1165 Soussana J-F, Graux A-I, Tubiello FN. Improving the use of modelling for projections of climate
1166 change impacts on crops and pastures. *Journal of Experimental Botany* 2010; 61: 2217-2228.
1167 doi: 10.1093/jxb/erq100

1168 Soussana J-F, Lemaire G. Coupling carbon and nitrogen cycles for environmentally sustainable
1169 intensification of grasslands and crop-livestock systems. *Agriculture, Ecosystems &*
1170 *Environment* 2014; 190: 9-17. doi: 10.1016/j.agee.2013.10.012

1171 Soussana JF, Loiseau P, Vuichard N, Ceschia E, Balesdent J, Chevallier T, et al. Carbon cycling and
1172 sequestration opportunities in temperate grasslands. *Soil Use and Management* 2004; 20:
1173 219-230. doi: 10.1111/j.1475-2743.2004.tb00362.x

1174 Steinauer K, Tilman D, Wragg PD, Cesarz S, Cowles JM, Pritsch K, et al. Plant diversity effects on soil
1175 microbial functions and enzymes are stronger than warming in a grassland experiment.
1176 *Ecology* 2015; 96: 99-112. doi: 10.1890/14-0088.1

1177 Suter M, Connolly J, Finn JA, Loges R, Kirwan L, Sebastià M-T, et al. Nitrogen yield advantage from
1178 grass-legume mixtures is robust over a wide range of legume proportions and
1179 environmental conditions. *Global Change Biology* 2015; 21: 2424-2438. doi:
1180 10.1111/gcb.12880

1181 Tardieu F, Simonneau T, Parent B. Modelling the coordination of the controls of stomatal aperture,
1182 transpiration, leaf growth, and abscisic acid: update and extension of the Tardieu-Davies
1183 model. *Journal of Experimental Botany* 2015. doi: 10.1093/jxb/erv039

1184 Thibault KM, Brown JH. Impact of an extreme climatic event on community assembly. *Proceedings of*
1185 *the National Academy of Sciences* 2008; 105: 3410-3415. doi: 10.1073/pnas.0712282105

1186 Thivierge M-N, Jégo G, Bélanger G, Bertrand A, Tremblay GF, Rotz CA, et al. Predicted Yield and
1187 Nutritive Value of an Alfalfa-Timothy Mixture under Climate Change and Elevated
1188 Atmospheric Carbon Dioxide. *Agronomy Journal* 2016; 108. doi: 10.2134/agronj2015.0484

1189 Thornton PK. Livestock production: recent trends, future prospects. *Philosophical Transactions:*
1190 *Biological Sciences* 2010; 365: 2853-2867. doi: 10.2307/20752983

1191 Thorsen SM, Höglind M. Modelling cold hardening and dehardening in timothy. Sensitivity analysis
1192 and Bayesian model comparison. *Agricultural and Forest Meteorology* 2010; 150: 1529-
1193 1542. doi: 10.1016/j.agrformet.2010.08.001

1194 Tixier P, Peyrard N, Aubertot J-N, Gaba S, Radoszycki J, Caron-Lormier G, et al. Modelling interaction
1195 networks for enhanced ecosystem services in agroecosystems. *Advances in Ecological*
1196 *Research* 2013; 49: 437-480. doi: 10.1016/B978-0-12-420002-9.00007-X

1197 Tylianakis JM, Didham RK, Bascompte J, Wardle DA. Global change and species interactions in
1198 terrestrial ecosystems. *Ecology Letters* 2008; 11: 1351-1363. doi: 10.1111/j.1461-
1199 0248.2008.01250.x

1200 Valqui Vidal RV. The future workshop: Democratic problem solving. IMM-Technical report-2005-7,
1201 Denmark, 2005.

1202 Van Oijen M, Höglind M. Toward a Bayesian procedure for using process-based models in plant
1203 breeding, with application to ideotype design. *Euphytica* 2015; 207: 627-643. doi:
1204 10.1007/s10681-015-1562-5

1205 Van Paassen A, Roetter RP, Van Keulen H, Hoanh CT. Can computer models stimulate learning about
1206 sustainable land use? Experience with LUPAS in the humid (sub-)tropics of Asia. *Agricultural*
1207 *Systems* 2007; 94: 874-887. doi: 10.1016/j.agsy.2006.11.012

1208 Venglovsky J, Sasakova N, Placha I. Pathogens and antibiotic residues in animal manures and
1209 hygienic and ecological risks related to subsequent land application. *Bioresource Technology*
1210 2009; 100: 5386-5391. doi: 10.1016/j.biortech.2009.03.068

1211 Verrelst J, Camps-Valls G, Muñoz-Marí J, Rivera JP, Veroustraete F, Clevers JGPW, et al. Optical
1212 remote sensing and the retrieval of terrestrial vegetation bio-geophysical properties – A

1213 review. *ISPRS Journal of Photogrammetry and Remote Sensing* 2015; 108: 273-290. doi:
 1214 10.1016/j.isprsjprs.2015.05.005
 1215 Vicca S, Serrano-Ortiz P, De Boeck HJ, Lemmens CMHM, Nijs I, Ceulemans R, et al. Effects of climate
 1216 warming and declining species richness in grassland model ecosystems: acclimation of CO₂
 1217 fluxes. *Biogeosciences Discussions* 2006; 3: 1473-1498. doi: 10.5194/bg-4-27-2007
 1218 Vital J-A, Gaurut M, Lardy R, Viovy N, Soussana J-F, Bellocchi G, et al. High-performance computing
 1219 for climate change impact studies with the Pasture Simulation model. *Computers and
 1220 Electronics in Agriculture* 2013; 98: 131-135. doi: 10.1016/j.compag.2013.08.004
 1221 Voinov A, Bousquet F. Modelling with stakeholders. *Environmental Modelling & Software* 2010; 25:
 1222 1268-1281. doi: 10.1016/j.envsoft.2010.03.007
 1223 Walz A, Braendle JM, Lang DJ, Brand F, Briner S, Elkin C, et al. Experience from downscaling IPCC-
 1224 SRES scenarios to specific national-level focus scenarios for ecosystem service management.
 1225 *Technological Forecasting and Social Change* 2014; 86: 21-32. doi:
 1226 10.1016/j.techfore.2013.08.014
 1227 Wheeler T, Reynolds C. Predicting the risks from climate change to forage and crop production for
 1228 animal feed. *Animal Frontiers* 2013; 3: 36-41. doi: 10.2527/af.2013-0006
 1229 Whish JPM, Herrmann NI, White NA, Moore AD, Kriticos DJ. Integrating pest population models with
 1230 biophysical crop models to better represent the farming system. *Environmental Modelling &
 1231 Software* 2015; 72: 418-425. doi: 10.1016/j.envsoft.2014.10.010
 1232 White SR, Carlyle CN, Fraser LH, Cahill JF. Climate change experiments in temperate grasslands:
 1233 synthesis and future directions. *Biology Letters* 2012; 8: 484-487. doi:
 1234 10.1098/rsbl.2011.0956
 1235 Wilkinson JM. Re-defining efficiency of feed use by livestock. *animal* 2011; 5: 1014-1022. doi:
 1236 10.1017/S175173111100005X
 1237 Wu L, McGechan MB, McRoberts N, Baddeley JA, Watson CA. SPACSYS: Integration of a 3D root
 1238 architecture component to carbon, nitrogen and water cycling-model description. *Ecological
 1239 Modelling* 2007; 200: 343-359. doi: 10.1016/j.ecolmodel.2006.08.010
 1240 Xi N, Carrère P, Bloor JMG. Nitrogen form and spatial pattern promote asynchrony in plant and soil
 1241 responses to nitrogen inputs in a temperate grassland. *Soil Biology and Biochemistry* 2014;
 1242 71: 40-47. doi: 10.1016/j.soilbio.2014.01.008
 1243 Yeluripati JB, del Prado A, Sanz-Cobeña A, Rees RM, Li C, Chadwick D, et al. Global Research Alliance
 1244 Modelling Platform (GRAMP): An open web platform for modelling greenhouse gas
 1245 emissions from agro-ecosystems. *Computers and Electronics in Agriculture* 2015; 111: 112-
 1246 120. doi: 10.1016/j.compag.2014.11.016
 1247 Yuan W, Cai W, Liu S, Dong W, Chen J, Arain MA, et al. Vegetation-specific model parameters are not
 1248 required for estimating gross primary production. *Ecological Modelling* 2014; 292: 1-10. doi:
 1249 10.1016/j.ecolmodel.2014.08.017
 1250 Zaka S, Frak E, Julier B, Gastal F, Louarn G. The thermal acclimation of photosynthesis only presents
 1251 limited intra-specific variations in a perennial crop selected over a broad climatic range. *AoB
 1252 Plants Accepted*. doi:
 1253 Zhao G, Siebert S, Enders A, Rezaei EE, Yan C, Ewert F. Demand for multi-scale weather data for
 1254 regional crop modeling. *Agricultural and Forest Meteorology* 2015; 200: 156-171. doi:
 1255 10.1016/j.agrformet.2014.09.026
 1256 Zhao S, Liu S. Scale criticality in estimating ecosystem carbon dynamics. *Global Change Biology* 2014;
 1257 20: 2240-2251. doi: 10.1111/gcb.12496
 1258 Zhao T, Ouyang Z, Jia L, Zheng H. Ecosystem services and their valuation of China grassland. *Acta
 1259 Ecologica Sinica* 2003; 24: 1101-1110. doi:
 1260 Zulka K, Götzl M. Ecosystem Services: Pest Control and Pollination. In: Steininger KW, König M,
 1261 Bednar-Friedl B, Kranzl L, Loibl W, Prettenhaler F, editors. *Economic Evaluation of Climate
 1262 Change Impacts*. Springer International Publishing, 2015, pp. 169-189.

1263 Zwicke M, Alessio GA, Thiery L, Falcimagne R, Baumont R, Rossignol N, et al. Lasting effects of
1264 climate disturbance on perennial grassland above-ground biomass production under two
1265 cutting frequencies. *Global Change Biology* 2013; 19: 3435-3448. doi: 10.1111/gcb.12317

1266

1267