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# Theory and the future of land-climate science

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1 **Climate over land—where humans live and the vast majority of food is produced—is chang-**  
2 **ing rapidly, driving severe impacts through extreme heat, wildfires, drought, and flooding.**  
3 **Our ability to monitor and model this changing climate is being transformed through new**  
4 **observational systems and increasingly complex Earth System Models (ESMs). But funda-**  
5 **mental understanding of the processes governing land climate has not kept pace, weakening**  
6 **our ability to interpret and utilise data from these advanced tools. Here we argue that for**  
7 **land-climate science to accelerate forward, an alternative approach is needed. We advocate**  
8 **for a parallel scientific effort, one emphasising robust theories, that aims to inspire current**  
9 **and future land-climate scientists to better comprehend the processes governing land climate,**  
10 **its variability and extremes, and its sensitivity to global warming. Such an effort, we believe,**  
11 **is essential to better understand the risks people face, where they live, in an era of climate**  
12 **change.**

13 Knowledge of some aspects of continental climate and their responses to global warming are  
14 well established. For example, we broadly understand why land warms more rapidly than oceans<sup>1</sup>  
15 (Fig. 1), the intensification of extreme precipitation in a warmer atmosphere<sup>2</sup>, and how surface  
16 runoff is influenced by loss of snowpack<sup>3</sup>. However, knowledge of many other aspects of land cli-  
17 mate is underdeveloped. The “wet get wetter, dry get drier” paradigm predicts an amplification of  
18 wet/dry contrasts as climate warms<sup>4</sup>. But this paradigm does not generally apply to land regions<sup>5</sup>  
19 nor does the poleward expansion of the Hadley cells<sup>6</sup>. Adding to this list is uncertainty over how  
20 evapotranspiration (ET) and soil moisture<sup>7,8</sup>—both critical for humans and ecosystems—will be  
21 altered by a changing climate. Knowledge of numerous other facets of land climate is similarly un-

22 settled, from basic questions of what governs its mean state, variability, and extremes, to how these  
23 facets might change with warming. Why are simulated land temperature changes more uncertain  
24 and more diverse, across space and climate models, compared to ocean regions (Fig. 1a,b)? Why  
25 are the tropical rainbelts broader and more mobile over land<sup>9</sup>? And how will land humidity evolve  
26 as climate warms<sup>10</sup>? Longstanding challenges in simulating land climate—including the diurnal  
27 cycle of convection<sup>11</sup>—further highlight shortcomings in our basic understanding.

### 28 **The challenge of complexity**

29 The climate over land is a complex system shaped by an array of diverse factors, from local surface  
30 conditions including soil moisture and plants<sup>12,13</sup> to large-scale atmospheric circulations that con-  
31 nect continents to oceans through the transport of water, heat, and momentum<sup>14,15</sup>. Many of the key  
32 processes influencing land climate are spatially heterogeneous, difficult to simulate, and/or poorly  
33 observed. For example, land surface models have longstanding problems in simulating turbulent  
34 fluxes of heat and water<sup>16,17</sup>, for reasons that are not well understood<sup>18</sup>. Sparse and time-limited  
35 observational records of important land-climate variables, including root-zone soil moisture<sup>19</sup> and  
36 near-surface humidity<sup>20</sup>, further impede efforts to advance knowledge of the land-climate system.  
37 The role of humanity presents another challenge, with large uncertainties in modelling the influ-  
38 ences of land use and management on fluxes of carbon, energy, and water in the past, present, and  
39 future<sup>21</sup>. Confronted with such a complex system it can appear a daunting task to develop a deep,  
40 mechanistic, conceptual understanding of the kind we would want to read in future textbooks on  
41 land climate. But as the field of climate science evolves, we argue that many of the most fascinating

42 and pressing questions relate to land.

43       Given the complexity and importance of land climate, how can the research community ac-  
44 celerate progress? In the atmospheric and ocean sciences, notable advances are being made by in-  
45 creasing the spatial resolution of state-of-the-art ESMs<sup>22</sup>. But unlike in the atmosphere and oceans,  
46 where higher resolutions allow for explicit simulation of key processes including deep convection  
47 and mesoscale eddies, the case for transitioning to finer resolution models to drive new conceptual  
48 breakthroughs in land-climate science is less clear-cut<sup>23</sup>. Land climate is undoubtedly influenced  
49 by small-scale processes, so there are potential benefits to incorporating into models more sophis-  
50 ticated representations of, for example, hillslope hydrology<sup>24</sup>, groundwater processes<sup>25</sup>, and land  
51 management<sup>26</sup>. However, complexity does not equate to realism: absent a comprehensive under-  
52 standing of these processes and how to accurately represent them in models<sup>27</sup>, it is possible that  
53 such complexity obfuscates more than it clarifies<sup>16</sup>. Persistent and poorly constrained deficiencies  
54 in land surface models—highlighted by the PLUMBER project<sup>16–18</sup>—suggest that model develop-  
55 ment alone, though vital, is unlikely to answer the key questions about land climate highlighted  
56 above. Similarly, machine learning tools are increasingly being applied to climate science for de-  
57 veloping ESMs<sup>28</sup>, parameterising surface fluxes<sup>29</sup>, and constructing statistical emulators of land  
58 models<sup>30</sup>. Indeed recent successes highlight the potential of machine learning to build physical  
59 insight in the atmospheric and ocean sciences<sup>31,32</sup>. But it remains to be seen whether the tools of  
60 machine learning are capable of transforming scientific understanding of land climate.

61 **A renewed focus on theory**

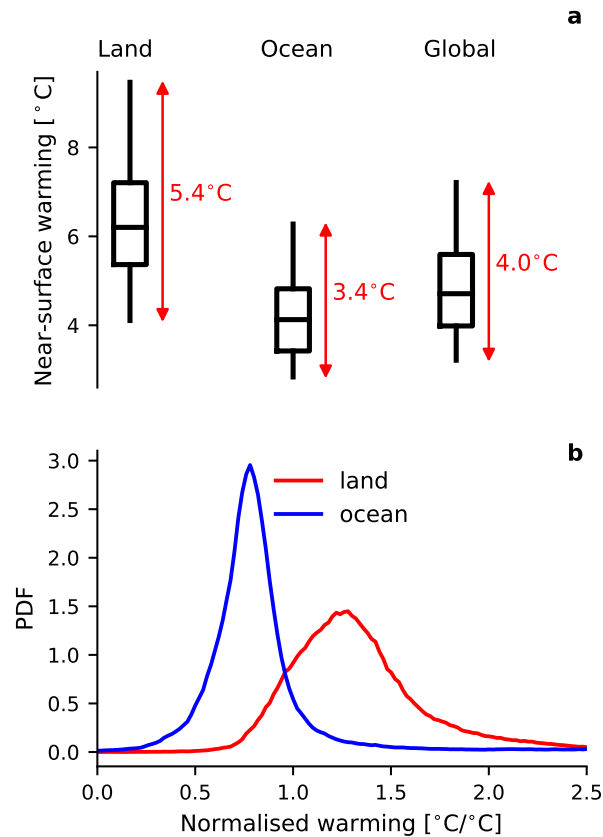


Figure 1: **Simulated climate warming is larger and more uncertain over land.** (a) Boxplots of simulated warming averaged over land (left), ocean (centre), and globally (right) calculated using pre-industrial control and abrupt 4xCO<sub>2</sub> simulations performed by 45 climate models participating in the Coupled Model Intercomparison Project Phase 6<sup>33</sup>. Horizontal lines show the median model values, boxes show the interquartile ranges, and whiskers show the full model ranges. Warming for each model is computed as the time- and area-averaged near-surface temperature change between the final 20 years of the pre-industrial control simulation and years 40-59 of the abrupt 4xCO<sub>2</sub> simulation. Uncertainty across models is indicated by the red arrows and text, with the full model range taken as a simple measure of uncertainty. (b) Multimodel-mean probability density functions (PDFs) of area-weighted near-surface warming over land (red) and ocean (blue), normalised by the global-mean warming in each model. The same<sup>6</sup> models, simulations, and averaging periods are used as in panel (a). The wider land PDF in panel (b) suggests larger differences in near-surface warming, across space and models, relative to oceans.

62 Here we argue that for land-climate science to move forward, we must step back and reassess our  
63 approach. Our philosophy—borne in an era of explosive growth in model complexity and demand-  
64 ing simulation timetables, and shaped by a 2022 workshop at the University of St Andrews—is to  
65 redouble efforts to build robust physical understanding of land climate through the development of  
66 powerful new theories and refinement of existing conceptual frameworks. Previous work exempli-  
67 fies this approach, notably the development of theories and simple ‘toy’ models to understand the  
68 land boundary layer<sup>34</sup>, land-atmosphere coupling<sup>35</sup>, and moist convection over land<sup>36</sup>. To anchor  
69 and inspire the next decade of research, we argue that now is the time to position this philosophy  
70 at the centre of land-climate science and re-balance our activities such that theory, model develop-  
71 ment, and observations are prioritised equally.

72 Development of theory can, and should, proceed in parallel with the imperative to build  
73 progressively more sophisticated ESMs. Indeed the gap in climate science between theory and  
74 actionable information, particularly at regional scales, is typically filled by state-of-the-art mod-  
75 els, which are also invaluable tools for testing and refining the theories advocated for here. But  
76 theories that distill conceptual understanding need to be at the core of land-climate science, to en-  
77 able the research community to compare proposed mechanisms, understand the competing roles of  
78 different processes in a coupled system, and make predictions without running complex models.  
79 Advances in theory can have practical as well as conceptual benefits, for example making ET easier  
80 to estimate<sup>37</sup>, increasing confidence in model projections (for example of runoff<sup>38</sup>), and underpin-  
81 ning physically-based emergent constraints to narrow uncertainties in future climate change<sup>39</sup>.



82 So, what constitutes a successful theory in land-climate science? The answer depends on  
83 the problem being considered, but we believe a successful theory should: explain an emergent  
84 property of the climate system; be underpinned by robust process understanding; and provide  
85 clear mechanistic insights that hold across a hierarchy of numerical model complexity. Theories  
86 should also, where possible, be predictive and quantitative (i.e., formulated as an equation or set  
87 of equations). Finally, and crucially, a successful theory should be tested against and supported by  
88 observational data. Below we highlight three recent advances in land-climate science that showcase  
89 the power of theory, before outlining our view on how a renewed focus on theory is needed to  
90 accelerate progress in land-climate science:

91 **1. Land temperature and humidity changes constrained by tropical atmospheric dynam-**

92 **ics:** The role of convection and large-scale atmospheric dynamics in shaping tropical land  
93 temperature and humidity has been an important conceptual advance over recent decades<sup>1,40,41</sup>.

94 This framework emerged from efforts to understand why, under climate change, warming is  
95 stronger over land; the so-called land-ocean warming contrast<sup>40</sup>. Early explanations of this  
96 phenomenon were based on the surface energy budget<sup>42</sup>. Radiative forcing at the surface  
97 (e.g., due to increases in atmospheric CO<sub>2</sub>) are largely balanced in ocean regions by in-  
98 creases in evaporation, resulting in a relatively small increase in surface temperature. In  
99 land regions, however, which are often water-limited, radiative forcing is primarily balanced  
100 through increases in sensible heat and longwave fluxes, requiring a larger increase in tem-  
101 perature relative to oceans. Though physically intuitive, using this argument to construct  
102 a quantitative theory for land temperature change is challenging because surface fluxes de-

103 pend on multiple factors aside from temperature, including windspeed, soil moisture, and  
104 the air-surface temperature disequilibrium.

105 An alternative framework, inspired by Joshi et al<sup>1</sup>, cuts through the complexity of land sur-  
106 faces to reveal a strong constraint on the response of tropical land to climate change. This  
107 framework has transformed understanding of the tropical land-ocean warming contrast and  
108 has led to broader insights into large-scale atmospheric controls on near-surface temperature  
109 and humidity. In the tropical atmosphere, strong vertical coupling by convection between the  
110 boundary layer and free troposphere described by convective quasi-equilibrium<sup>43</sup>—together  
111 with horizontal coupling by gravity waves above the boundary layer, resulting in weak  
112 free-tropospheric temperature gradients<sup>44</sup>—imply that climatic changes in adiabatically con-  
113 served quantities such as moist static energy, a function of temperature and specific humidity  
114 near the surface, are tightly coupled between different regions and therefore approximately  
115 uniform on large scales<sup>45–47</sup> (Fig. 2). This mechanism, a form of ‘downward control’ ex-  
116 erted by the overlying atmosphere on near-surface tropical climate, has important implica-  
117 tions: Though temperature and specific humidity individually may respond differently to  
118 climate change in different regions, for example in tropical savannas versus in rainforests,  
119 the combined change (encoded in the moist static energy) is more spatially homogeneous.  
120 Local processes, including soil moisture and aridity<sup>46,48</sup>, are crucial for controlling how tem-  
121 perature versus humidity changes contribute to the change in moist static energy imposed  
122 by the atmosphere. This physical theory underpins advances in understanding the land-  
123 ocean warming contrast<sup>1,49</sup>, aridity and land relative humidity in a changing climate<sup>41,46,50</sup>,

124 and extreme heat<sup>47,51,52</sup>, and establishes a simple yet quantitative framework for interpreting  
125 models, observations, and the roles of local versus large-scale processes in shaping tropical  
126 land climate.

127 **2. Evapotranspiration predicted by simple theory:** ET is central to regulating the water, en-  
128 ergy, and carbon budgets of land regions<sup>53</sup>, and affects societies and ecosystems through its  
129 influence on hydrology and temperature variability<sup>54</sup>. But ET is directly measured only at a  
130 limited number of sites<sup>55</sup>, necessitating models of various kinds to estimate ET elsewhere.  
131 These models are typically complex, requiring numerous poorly constrained land-surface  
132 parameters as inputs, and are imperfect at replicating direct measurements<sup>56</sup>. However, a  
133 new theory to predict present-day ET in inland continental regions using minimal input data  
134 provides a conceptual advance in understanding and presents an opportunity to greatly ex-  
135 pand the database of ET measurements across space and time<sup>37</sup>. The theory is based on the  
136 concept of ‘surface flux equilibrium’ (SFE), which assumes an approximate balance between  
137 the surface moistening and heating effects on near-surface relative humidity<sup>57</sup>. This strong  
138 coupling between the land surface and overlying atmosphere imprints, in the air properties,  
139 information about the land-surface fluxes (i.e., the Bowen ratio) at daily to longer timescales,  
140 and appears to dominate alternative atmospheric mechanisms that also contribute to deter-  
141 mining the near-surface atmospheric state (e.g., wind-driven moisture and heat convergence).  
142 Specifically, the SFE theory permits relatively accurate estimates of ET knowing only the net  
143 radiative flux into the surface and the near-surface temperature and specific humidity<sup>37,58</sup>,  
144 the latter two which reflect the Bowen ratio (Fig. 3). Importantly, these quantities are more

145 widely available from weather stations than direct ET measurements. The theory reveals an  
146 emergent simplicity in ET<sup>37</sup>, despite the heterogeneity and complexity of land surfaces.

147 **3. Leaf physiology incorporated into classical runoff theories:** Runoff from land supplies  
148 almost all the water used by humans. In contrast to the time-varying ET estimated by SFE  
149 and described above, long-term mean runoff and ET fluxes have long been predicted and  
150 understood using the simple theory of Budyko<sup>59</sup>, in which the fraction of precipitation that  
151 becomes runoff decreases as the ratio of atmospheric evaporative demand to precipitation  
152 increases. Budyko quantified evaporative demand using surface net radiation only, but more  
153 comprehensive evaporative theories<sup>60</sup> generally also include a well-understood positive tem-  
154 perature dependence<sup>61</sup>. When these more modern methods are used in the Budyko theory,  
155 they predict substantial increases in evaporative demand with global warming and systematic  
156 decreases in natural runoff<sup>62</sup> (i.e., the component of runoff controlled by natural processes  
157 rather than by human activities), which would imply water shortages. Yet such widespread  
158 runoff declines are neither observed<sup>63</sup> nor simulated by more comprehensive models<sup>62</sup>, lead-  
159 ing to the impression of a theoretical deficiency. Yang et al<sup>64</sup> recently resolved this tension  
160 by incorporating the ET-reducing closure of leaf stomata by CO<sub>2</sub> into a revised theoret-  
161 ical framework (Fig. 4). The inclusion of this important and well-studied process brought  
162 the Budyko-predicted trends in natural runoff much closer to observations and state-of-the-  
163 art ESMs, and clarified our understanding of the drivers of runoff in a changing climate.  
164 Looking forward, incorporating human activities (e.g., water management) and the effects  
165 of wildfire into runoff theories is a priority for future work.

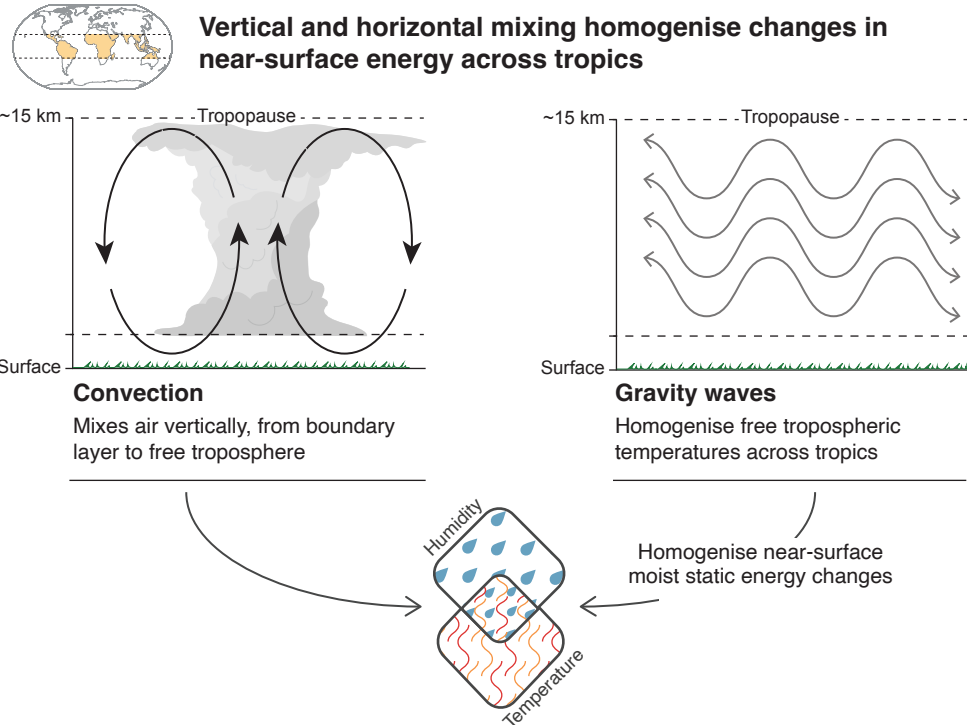


Figure 2: **Atmospheric dynamics constrain changes in tropical land climate.** Schematic illustrating how convection and gravity waves in the tropical atmosphere spatially homogenise climatic changes in near-surface moist static energy. The development of this large-scale atmospheric constraint on tropical land climate has been an important conceptual advance over recent years. Here and in Figures 3 and 4, the title maps highlight where the mechanism is broadly expected to be applicable.

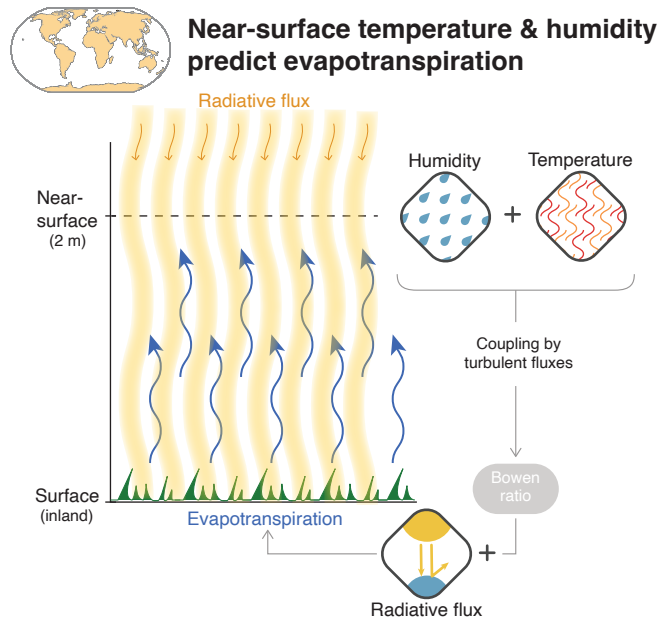


Figure 3: **Evapotranspiration inferred from temperature and humidity measurements.** Schematic highlighting how, following recent theoretical developments, inland ET can be predicted as a simple function of near-surface temperature and humidity along with the net radiative flux into the surface. Note that the grey arrows represent the series of inferences used by the SFE-based theory to make estimates of  $ET^{37}$ , whereas the blue and orange arrows denote, respectively, the turbulent fluxes of heat and water coupling the surface to the near-surface air and the radiative energy fluxes.

 **As CO<sub>2</sub> rises, plants limit evapo-  
transpiration and boost river runoff**

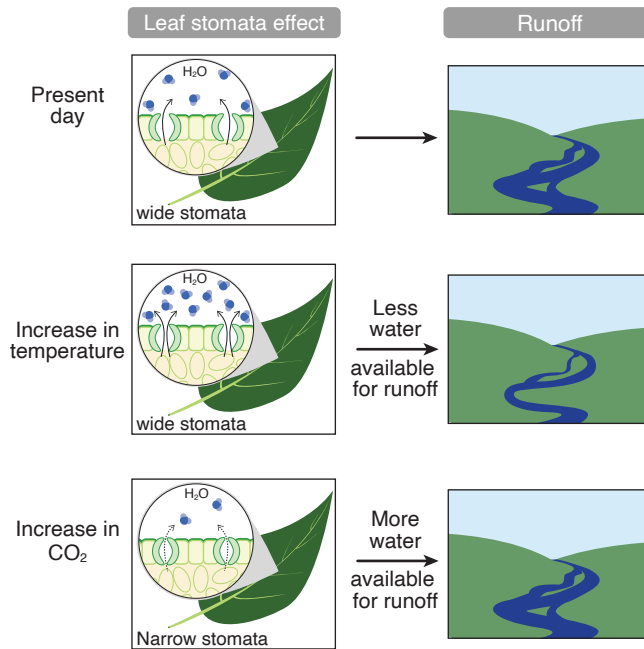


Figure 4: **Stomatal response to increasing CO<sub>2</sub> boosts river runoff.** Schematic depicting the competing effects of temperature versus CO<sub>2</sub> on ET from leaves and on river runoff. The recent incorporation of the CO<sub>2</sub> effect into classical theories has clarified understanding of runoff in a changing climate.

166 **Opportunities for progress**

167 A greater emphasis on developing theories for land climate and its changes is essential for build-  
168 ing confidence in future projections, identifying directions for model improvement, validating *in*  
169 *situ* and remote sensing data, and interpreting the dynamics of key processes as new models and  
170 observational systems come online. The examples highlighted above demonstrate the potential for  
171 theory to further fundamental understanding of land climate. But the next set of advances is now  
172 needed. Below we present three areas of land-climate science primed for theory to provide new  
173 insights:

174 **1. Atmospheric circulation and land:** The atmospheric circulation strongly shapes the land  
175 climate, from extreme temperatures<sup>65</sup> to the regional water cycle<sup>66</sup>. However, much of our  
176 understanding of the atmospheric circulation and its sensitivity to climate change has been  
177 developed using aquaplanet models without land surfaces<sup>67,68</sup>. Over recent years, focus has  
178 begun to shift towards incorporating land into conceptual frameworks for the atmospheric  
179 state and circulation<sup>69-71</sup>. But numerous basic questions persist, including: Why is the tropi-  
180 cal rainbelt wider over continents<sup>9</sup>? How can ingredients of the land surface be incorporated  
181 into modern theories for monsoons<sup>72</sup>? Why is the poleward expansion of the atmospheric  
182 circulation under global warming much weaker over land<sup>6</sup>? How will blocks, often the cause  
183 of extreme weather over land, change with warming<sup>73</sup>? And what processes control updraft  
184 velocities—and hence influence extreme precipitation—over land<sup>2</sup>? These important ques-  
185 tions are ready to be tackled with novel theories.



186 2. **Water and land:** Beyond a broad tendency for mean relative humidity over land to decrease  
187 with warming<sup>41,50,74</sup>, basic properties of the land water cycle and its response to climate  
188 change remain unexplained. For example, what are the mechanisms determining the spatial  
189 and temporal distribution of soil moisture in the current climate<sup>75</sup>? Why do climate mod-  
190 els project drier surface soils in most regions<sup>8</sup>? And why do future trajectories for surface  
191 and column soil moisture differ<sup>76</sup>? Detailed understanding of near-surface humidity over  
192 land is another priority<sup>10</sup>, given the strong coupling to trends in extreme temperatures<sup>52,77</sup>,  
193 extreme precipitation<sup>78</sup>, and runoff<sup>79</sup>. The coupling between plants and water has major  
194 implications for drought and terrestrial ecosystems, yet its response to climate change is  
195 highly uncertain<sup>80</sup>. For example, the effects of plant changes on runoff beyond the simple  
196 CO<sub>2</sub>-stomatal dependence<sup>64</sup> are likely very large<sup>81</sup> but poorly understood. Finally the phe-  
197 nomenon of ‘flash droughts’, whose dynamics and predictability are only beginning to be  
198 explored<sup>82</sup>, is an emerging topic where creative new theories are needed.

199 3. **Carbon and land:** Carbon uptake and release by terrestrial ecosystems both affects and re-  
200 sponds to climate variability and long-term change. The field of carbon-water-climate feed-  
201 backs is already rich with examples of simple concepts, theories, and emergent constraints<sup>83–85</sup>,  
202 providing a way to synthesise or contrast the behaviours emerging from complex ESMs<sup>86</sup>.  
203 The carbon-concentration and carbon-climate feedback parameters, for example, encapsu-  
204 late the overall response of land carbon stocks to changes in atmospheric CO<sub>2</sub> and to global  
205 warming, respectively<sup>87</sup>. This global-scale conceptual framework can be used to diagnose  
206 and compare complex simulations<sup>88</sup>, but is also transferable to climate emulators or models

207 of reduced complexity<sup>89</sup>. However, similarly simple and adaptable concepts are lacking in  
208 other areas of carbon cycle research. There is, for instance, large uncertainty on the extent to  
209 which tipping points at regional scales could impact some of the world's largest carbon pools,  
210 like permafrost carbon, the Amazon rainforest ecosystem, and global forests<sup>90-93</sup>. To some  
211 extent this is because we lack theories, metrics, and frameworks to explain and reconcile the  
212 contradicting results obtained from different models and approaches. However, the existing  
213 literature on dynamical systems theory is rich with concepts that may be transferable to un-  
214 derstand potential tipping points in the carbon cycle if they can be adequately constrained by  
215 observations, similar to what has been done to study transitions between stable system states  
216 or attractors in ecology and population dynamics<sup>94,95</sup>.

## 217 **Outlook**

218 To discover, test, and refine the powerful theories for land climate advocated for in this perspective,  
219 and to maximise benefits for the wider climate community, technical tools and scientific talent are  
220 needed. On the tools side, we have at our disposal a range of models spanning idealised<sup>96</sup> to state-  
221 of-the-art ESMs<sup>33</sup>, alongside the emerging generation of 'global storm resolving' models<sup>22</sup> and  
222 flexible, process-based hydrologic models<sup>97</sup>. This model hierarchy is well positioned for building  
223 new understanding of land climate. However, a lack of observations presents a major challenge<sup>98</sup>:  
224 Despite recent progress, for example in remote sensing of surface soil moisture<sup>99</sup>, we simply do  
225 not have long-term datasets with wide spatial coverage for many important land-climate quantities,  
226 including root-zone soil moisture and ET. Thus, to parallel the development of models and efforts

227 to construct theories for land climate, new instrumental observations of essential land surface fluxes  
228 and reservoirs are required. Opportunities to further leverage existing observational datasets, with  
229 the goal of improving models and testing theories, should also be exploited. Beyond observational  
230 uncertainty, whenever we ground new theory in observations we also have to contend with the  
231 complicating influence of internal climate variability. Separating the forced response from internal  
232 variability at regional scales is still challenging and can harbour surprises that can influence our  
233 theories<sup>100</sup>. Empirical-statistical methods to isolate the forced response, and new theory on internal  
234 variability itself, will thus need to accompany our endeavour to refine understanding of land climate  
235 and its changes with warming.

236 On the talent side, to tackle the important questions in land-climate science we need to con-  
237 tinually inspire, recruit, and resource diverse cohorts of researchers from a range of primary disci-  
238 plines spanning atmospheric science, hydrology, ecology, physics, mathematics, computer science,  
239 and beyond. Engaging scientists from the broader climate community—those working primar-  
240 ily on atmospheric dynamics, for example—also has the potential to bring new ideas and drive  
241 progress in land-climate science. Through this perspective, alongside a series of workshops and  
242 summer schools we aim to coordinate over coming years, our goal is to engage these current and  
243 future generations of researchers—as well as major funding bodies and established land-focused  
244 research initiatives—in our vision to place theory at the core of land-climate science.

245 State-of-the-art models, observational systems, and machine learning are transforming our  
246 ability to simulate, monitor, and emulate many aspects of land climate. But our scientific under-

247 standing has not kept pace, and we now lack robust theories to comprehend the rich complexity  
248 being revealed by these advanced tools. Now is the time to change course and underpin models,  
249 observations, and machine-learning techniques with new theories so that we maintain and advance  
250 the deep, mechanistic understanding of land climate needed to meet the challenges of an uncertain  
251 future.

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498 **Code Availability** The code used to produce Figure 1 is available from the corresponding author on re-  
499 quest.