

# A simple explanation for declining temperature sensitivity with warming

E. M. Wolkovich<sup>1,a</sup>, J. L. Auerbach<sup>2</sup>, C. J. Chamberlain<sup>3</sup>, D. M. Buonaiuto<sup>3</sup>,  
A. K. Ettinger<sup>4</sup>, I. Morales-Castilla<sup>5</sup> & A. Gelman<sup>2</sup>

<sup>1</sup>Forest & Conservation Sciences, Faculty of Forestry, University of British Columbia, Vancouver, British Columbia, Canada

<sup>2</sup>Department of Statistics, Columbia University, New York, NY 10027, USA

<sup>3</sup>Department of Organismic and Evolutionary Biology, Harvard University, Cambridge, Massachusetts, USA

<sup>4</sup>The Nature Conservancy in Washington, 74 Wall Street, Seattle, WA, USA

<sup>5</sup>Department of Life Sciences, University of Alcalá CTRA N-II, KM., 33,600, 28802, Alcalá de Henares, Spain

<sup>a</sup>Corresponding author (ORCID: 0000-0001-7653-893X)

## Abstract

1  
2       Temperature sensitivity—the magnitude of a biological response per °C—is a fundamen-  
3       tal concept across scientific disciplines, especially biology, where temperature determines the  
4       rate of many plant, animal and ecosystem processes. Recently, a growing body of literature  
5       in global change biology has found temperature sensitivities decline as temperatures rise (Fu  
6       et al., 2015; Güsewell et al., 2017; Piao et al., 2017; Dai et al., 2019). Such observations have  
7       been used to suggest climate change is reshaping biological processes, with major implications  
8       for forecasts of future change. Here we present a simple alternative explanation for observed  
9       declining sensitivities: the use of linear models to estimate non-linear temperature responses.  
10       Corrections for the non-linearity of temperature response in simulated data and long-term  
11       phenological data from Europe remove the apparent decline. Our results show that rising  
12       temperatures combined with linear estimates based on calendar time produce observations  
13       of declining sensitivity—without any shift in the underlying biology. Current methods may  
14       thus undermine efforts to identify when and how warming will reshape biological processes.

# 1 Main text

Climate change has reshaped biological processes around the globe, with shifts in the timing of major life history events (phenology), carbon dynamics and other ecosystem processes (IPCC, 2014). With rising temperatures, a growing body of literature has documented changes in temperature sensitivity—the magnitude of a biological response scaled per °C. Many studies have found declining responses to temperature in recent decades (Fu et al., 2015; Güsewell et al., 2017; Piao et al., 2017; Dai et al., 2019), and some have reported more uniform sensitivities across elevation (Vitasse et al., 2018), or lower sensitivities in warmer, urban areas (Meng et al., 2020).

Most studies attribute changes in temperature sensitivity to shifts in underlying biological processes. For example, researchers have suggested weaker temperature sensitivities are evidence of increased light limitation in the tundra (Piao et al., 2017), or a decline in the relative importance of warm spring temperatures for spring phenological events (e.g., leafout, insect emergence) in the temperate zone (Fu et al., 2015; Meng et al., 2020), as other environmental triggers (e.g., winter temperatures that determine ‘chilling’) play a larger role. Yet, despite an increase in studies reporting declining or shifting temperature sensitivities, none have provided strong evidence of the biological mechanisms underlying these changes (e.g., Fu et al., 2015; Meng et al., 2020). The missing mechanisms may be hidden in the data: environmental factors moderate biological processes in complex ways (Chuine et al., 2016; Güsewell et al., 2017), are strongly correlated in nature (e.g., Fu et al., 2015), and temperature variance shifts over time and space (Keenan et al., 2020).

Here we propose a simpler alternative explanation: the use of linear models for non-linear responses to temperature. Researchers generally use methods with assumptions of linearity to calculate temperature sensitivities, often relying on some form of linear regression to compute a change in a quantity—days to leafout or carbon sequestered over a fixed time, for example—per °C, thus ignoring that many biological responses to temperature are non-linear. We show, theoretically then with simulated and empirical data, how the use of linear methods for non-

44 linear responses can produce an illusion that the mechanisms underlying biological processes are  
45 changing.

46

47 Many observed biological responses are the result of continuous non-linear processes that depend  
48 on temperature, which are discretized into temporal units for measurement. For example, a bio-  
49 logical response, such as leafout, occurs when a certain thermal sum is reached (Dijkhuis, 1956;  
50 Lindsey and Newman, 1956), and plants will reach this threshold more quickly—in calendar  
51 time—when average daily temperatures are warmer (Valentine, 1983; Lechowicz, 1984; Kramer,  
52 2012). Biologically, however, the plants may require the same temperature sum. Indeed any  
53 process observed or measured as the time until reaching a threshold is inversely proportional to  
54 the speed at which that threshold is approached. Temperature determines the speed of many  
55 biological processes (Bonan and Sirois, 1992; Hinrichsen, 2009; Hofmann and Todgham, 2010).  
56 Thus, at very low temperatures plants would never leaf out and at higher temperatures they  
57 could leaf out in only a matter of days—yet sensitivities estimated from linear regression at  
58 higher (warmer) temperatures would appear much lower than those observed at lower temper-  
59 atures. Warming acts to step on the biological accelerator, producing shifts in estimates when  
60 non-linear responses are modeled as linear.

61

62 We show this by deriving the relationship between a biological response and temperature using  
63 a simple stochastic model, which describes the first time a random process hits a threshold (see  
64 ‘A first-hitting-time model of leafout’ in Supplementary Information). Our model holds the  
65 temperature threshold for leafout constant (Hunter and Lechowicz, 1992; Man and Lu, 2010;  
66 Zohner et al., 2020). Even though the mechanism by which temperature leads to leafout does not  
67 change, the model produces declining sensitivity—as measured in days per °C—with warming.  
68 Indeed, under this model constant temperature sensitivity would be evidence that the temper-  
69 ature threshold is not constant and the mechanisms underlying the leafout process have changed.

70

71 Simulations show that correcting for non-linearity using a log transformation removes appar-  
72 ent declines in temperature sensitivity (Fig. 1, S2, code link). In empirical long-term leafout  
73 data from Europe, correcting for non-linearity in responses produces little evidence for declining

74 sensitivities with warming (Figs. 1, S6, S7). An apparent decline in sensitivity for silver birch  
75 (*Betula pendula*) from  $-4.3$  days/ $^{\circ}\text{C}$  to  $-3.6$  days/ $^{\circ}\text{C}$  from 1950-1960 compared to 2000-2010 dis-  
76 appears using a log-log regression ( $-0.17$  versus  $-0.22$ ). We see similar corrections using 20-year  
77 windows, and a potential increase in sensitivity for European beech (*Fagus sylvatica*, see Ta-  
78 bles S1-S2). Moreover, the variance of the leafout dates of both species declines as temperatures  
79 rise—(declines of roughly 50%, see Tables S1-S2), which is expected under our model as warming  
80 accelerates towards the thermal threshold that triggers leafout (and in contrast to predictions  
81 from changing mechanisms, see Ford et al., 2016).

82

83 Fundamentally rising temperature should alter many biological processes, making robust meth-  
84 ods for identifying these changes critical. In spring plant phenology, where declining sensitivities  
85 are often reported (Fu et al., 2015; Piao et al., 2017; Dai et al., 2019), warming may increase  
86 the role of ‘chilling’ (determined mainly by winter temperatures) and daylength (Laube et al.,  
87 2014; Zohner et al., 2016)—potentially increasing the thermal sum required for leafout at lower  
88 values of these cues (Polgar et al., 2014; Zohner et al., 2017; Flynn and Wolkovich, 2018). Ad-  
89 justing our simulations to match this model yielded shifts in sensitivities with warming. Unlike  
90 a model with no underlying biological change, however, after correcting for non-linearity, the  
91 shifts in sensitivities remained and they occurred in step with the biological change (Fig. S4a,  
92 c). In contrast, sensitivities estimated from a linear model showed shifts across the entire range  
93 of warming, well before the simulated biological change (Fig. S4a, c). Further, we found that  
94 an increase in the thermal sum required for leafout should yield larger in magnitude temper-  
95 ature sensitivities, not smaller, as often expected (e.g., Fu et al., 2015), thus highlighting the  
96 complexity of what trends to expect in sensitivities with warming (see ‘Common hypotheses for  
97 declining sensitivity’ in Supplementary Information for an extended discussion).

98

99 Our theoretical model and empirical results show that rising temperatures are sufficient to ex-  
100 plain declining temperature sensitivity. It is not necessary to invoke changes to the mechanisms  
101 that underlie the biological processes themselves. Our results provide a simpler explanation for  
102 observations of declining temperature sensitivities, but do not rule out that important changes  
103 in biological processes may underlie such declines. Instead, our results highlight how the use

104 of linear models may make identifying when—and why—warming alters underlying biology far  
105 more difficult.

106

107 Inferring biological processes from statistical artifacts is not a new problem (e.g., Nee et al.,  
108 2005), but climate change provides a new challenge in discerning mechanism from measure-  
109 ments because it affects biological time, while researchers continue to use calendar time. Other  
110 fields focused on temperature sensitivity often use approaches that acknowledge the non-linearity  
111 of responses (e.g., Yuste et al., 2004). Researchers have called for greater use of process-based  
112 models (Keenan et al., 2020), which often include non-linear responses to temperature, but  
113 rely themselves on exploratory methods and descriptive analyses for progress (Chuine et al.,  
114 2016). The challenge, then, is to interrogate the implicit and explicit models we use to interpret  
115 data summaries, and to develop null expectations that apply across biological and calendar time.

116

## 117 **References**

118 Bonan, G. B., and L. Sirois. 1992. Air-temperature, tree growth, and the northern and southern  
119 range limits to *Picea-mariana*. *Journal of Vegetation Science* 3:495–506.

120 Chuine, I., M. Bonhomme, J.-M. Legave, I. García de Cortázar-Atauri, G. Charrier, A. Lacointe,  
121 and T. Améglio. 2016. Can phenological models predict tree phenology accurately in the  
122 future? The unrevealed hurdle of endodormancy break. *Global Change Biology* 22:3444–  
123 3460.

124 Dai, W. J., H. Y. Jin, Y. H. Zhang, T. Liu, and Z. Q. Zhou. 2019. Detecting temporal changes in  
125 the temperature sensitivity of spring phenology with global warming: Application of machine  
126 learning in phenological model. *Agricultural and Forest Meteorology* 279.

127 Dijkhuis, F. J. 1956. Computation of heat unit accumulations in maize for practical application.  
128 *Euphytica* 5:267–275.

129 Flynn, D. F. B., and E. M. Wolkovich. 2018. Temperature and photoperiod drive spring phe-  
130 nology across all species in a temperate forest community. *New Phytologist* 219:1353–1362.

- 131 Ford, K. R., C. A. Harrington, S. Bansal, J. Gould, Peter, and J. B. St. Clair. 2016. Will changes  
132 in phenology track climate change? A study of growth initiation timing in coast Douglas-fir.  
133 *Global Change Biology* 22:3712–3723.
- 134 Fu, Y. S. H., H. F. Zhao, S. L. Piao, M. Peaucelle, S. S. Peng, G. Y. Zhou, P. Ciais, M. T. Huang,  
135 A. Menzel, J. P. Uelas, Y. Song, Y. Vitasse, Z. Z. Zeng, and I. A. Janssens. 2015. Declining  
136 global warming effects on the phenology of spring leaf unfolding. *Nature* 526:104–107.
- 137 Güsewell, S., R. Furrer, R. Gehrig, and B. Pietragalla. 2017. Changes in temperature sensitivity  
138 of spring phenology with recent climate warming in Switzerland are related to shifts of the  
139 preseason. *Global Change Biology* 23:5189–5202.
- 140 Hinrichsen, H. H. 2009. Biological processes and links to the physics. *Deep-Sea Research Part*  
141 *ii-Topical Studies in Oceanography* 56:1968–1983.
- 142 Hofmann, G. E., and A. E. Todgham. 2010. Living in the now: Physiological mechanisms to  
143 tolerate a rapidly changing environment. *Annual Review of Physiology* 72:127–145.
- 144 Hunter, A. F., and M. J. Lechowicz. 1992. Predicting the timing of budburst in temperate trees.  
145 *Journal of Applied Ecology* 29:597–604.
- 146 IPCC. 2014. *Climate Change 2014: Impacts, Adaptation, and Vulnerability*. Cambridge Uni-  
147 versity Press, Cambridge, United Kingdom and New York, NY, USA.
- 148 Keenan, T. F., A. D. Richardson, and K. Hufkens. 2020. On quantifying the apparent temper-  
149 ature sensitivity of plant phenology. *New Phytologist* 225:1033–1040.
- 150 Kramer, P. 2012. *Physiology of woody plants*. Elsevier, New York.
- 151 Laube, J., T. H. Sparks, N. Estrella, J. Höfler, D. P. Ankerst, and A. Menzel. 2014. Chilling  
152 outweighs photoperiod in preventing precocious spring development. *Global Change Biology*  
153 20:170–182.
- 154 Lechowicz, M. J. 1984. Why do temperate deciduous trees leaf out at different times - adaptation  
155 and ecology of forest communities. *American Naturalist* 124:821–842. Lechowicz, mj.

- 156 Lindsey, A. A., and J. E. Newman. 1956. Use of official weather data in spring time - temperature  
157 analysis of an Indiana phenological record. *Ecology* 37:812–823.
- 158 Man, R. Z., and P. X. Lu. 2010. Effects of thermal model and base temperature on estimates of  
159 thermal time to bud break in white spruce seedlings. *Canadian Journal of Forest Research-*  
160 *Revue Canadienne De Recherche Forestiere* 40:1815–1820.
- 161 Meng, L., J. Mao, Y. Zhou, A. D. Richardson, X. Lee, P. E. Thornton, D. M. Ricciuto, X. Li,  
162 Y. Dai, X. Shi, and G. Jia. 2020. Urban warming advances spring phenology but reduces the  
163 response of phenology to temperature in the conterminous United States. *Proceedings of the*  
164 *National Academy of Sciences* 117:4228.
- 165 Nee, S., N. Colegrave, S. A. West, and A. Grafen. 2005. The illusion of invariant quantities in  
166 life histories. *Science* 309:1236–1239.
- 167 Piao, S., Z. Liu, T. Wang, S. Peng, P. Ciais, M. Huang, A. Ahlstrom, J. F. Burkhart, F. Cheval-  
168 lier, I. A. Janssens, et al. 2017. Weakening temperature control on the interannual variations  
169 of spring carbon uptake across northern lands. *Nature Climate Change* 7:359.
- 170 Polgar, C., A. Gallinat, and R. B. Primack. 2014. Drivers of leaf-out phenology and their  
171 implications for species invasions: insights from Thoreau’s Concord. *New Phytologist* 202:106–  
172 15.
- 173 Valentine, H. T. 1983. Budbreak and leaf growth functions for modeling herbivory in some  
174 gypsy-moth hosts. *Forest Science* 29:607–617.
- 175 Vitasse, Y., C. Signarbieux, and Y. H. Fu. 2018. Global warming leads to more uniform spring  
176 phenology across elevations. *Proceedings of the National Academy of Sciences* 115:1004–1008.
- 177 Yuste, J., I. A. Janssens, A. Carrara, and R. Ceulemans. 2004. Annual  $Q_{10}$  of soil respiration  
178 reflects plant phenological patterns as well as temperature sensitivity. *Global Change Biology*  
179 10:161–169.
- 180 Zohner, C. M., B. M. Benito, J. D. Fridley, J. C. Svenning, and S. S. Renner. 2017. Spring  
181 predictability explains different leaf-out strategies in the woody floras of North America,  
182 Europe and East Asia. *Ecology Letters* 20:452–460.



183 Zohner, C. M., B. M. Benito, J. C. Svenning, and S. S. Renner. 2016. Day length unlikely to  
184 constrain climate-driven shifts in leaf-out times of northern woody plants. *Nature Climate*  
185 *Change* 6:1120–1123.

186 Zohner, C. M., L. D. Mo, T. A. M. Pugh, J. F. Bastin, and T. W. Crowther. 2020. Interactive  
187 climate factors restrict future increases in spring productivity of temperate and boreal trees.  
188 *Global Change Biology* 26:4042–4055.

189 *Acknowledgements:* Thanks to TJ Davies, A Donnelly, TM Giants, D. Lipson, C. Rollinson and  
190 C. Zohner.

191

192 *Data & Code Availability:* Code for simulations and plots is provided here. For empirical ex-  
193 amples, we used PEP 725 phenological data and E-OBS climate data, both of which are freely  
194 available via the links.

195

196 *List of Supplementary Information:*

197 A first-hitting-time model of leafout

198 Simulations of common hypotheses for declining sensitivity

199 Methods & results using long-term empirical data (PEP725)

200 Table S1-S2

201 Fig S1-S7

202

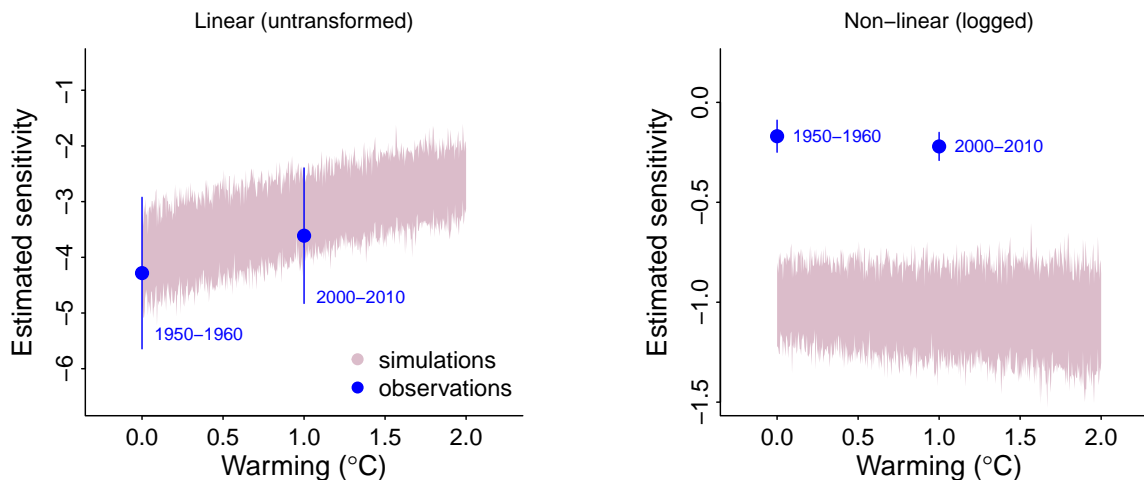


Figure 1: **Shifts in temperature sensitivities (response per °C) with warming occur when using linear models for non-linear processes.** Estimated sensitivities decline (in magnitude) with warming in simulations (shading, estimated across 45 sites with a base temperature of normal(6,4), variation comes from fluctuation in the Monte Carlo simulations) with no underlying change in the biological process when sensitivities were estimated with linear regression (left). This decline disappears when performing the regression on logged predictor and response variables (right). Such issues may underlie declining sensitivities calculated from observational data, including long-term observations of leafout across Europe (for *Betula pendula* from PEP725 from for the 45 sites that had complete data for 1950-1960 and 2000-2010), which show a lower sensitivity with warming when calculated on raw data, but no change in sensitivity using logged data. Shading, symbols and lines represent means  $\pm$  standard deviations of regressions across sites. See Supplementary Information for a discussion of why estimated sensitivities are -1 or lower in non-linear models.