Supplemental materials: How environmental tracking shapes species and communities in stationary and non-stationary systems

E. M. Wolkovich & M. J. Donahue

1 Literature review of studies examining tracking & other traits

To examine current evidence of what traits may trade-off with tracking, we systematically reviewed the literature for studies examining tracking and other traits together. We searched ISI in August 2019 for:

- 1. Topic: 'phenolog* chang*' and Title: phenolog* AND trait*
- 2. Topic: 'warming shift*' AND trait* and Title: phenolog*
- 3. Topic: 'phenolog* track*' AND trait* and Title: phenolog*
- 4. Topic: 'phenolog* sensitiv*' AND trait* and Title: phenolog*

which resulted in 231 papers (83% of which were published in 2011 or later, see Fig. S1). From here we used the following criteria to determine from which papers we could not extract data: no phenology or phenological change measured (73 papers), no trait(s) measured or analyzed (49 papers), single-species studies focused on intra-specific variation (55 papers), modeling or theory studies without data (12 papers), or papers without new data presented (reviews, etc.: 4 papers), or miscellaneous reasons leading to no data relevant to our aims (7 papers). This left us with 30 papers including relevant data (Suzuki & Kudo, 1997; Post & Stenseth, 1999; Adrian et al., 2006; Xu et al., 2009; Goodenough et al., 2010; Diamond et al., 2011; Moussus et al., 2011; Szilvia et al., 2012; Dorji et al., 2013; Ishioka et al., 2013; Xia & Wan, 2013; Bock et al., 2014; Kharouba et al., 2014; Vegvari et al., 2015; Bell et al., 2015; Jing et al., 2016; Lasky et al., 2016; McDermott & DeGroote, 2016; Zhu et al., 2016; Brooks et al., 2017; Du et al., 2017; Munson & Long, 2017; Arfin Khan et al., 2018; Zhang et al., 2018; Ladwig et al., 2019; Park et al., 2019; Sharma & Upadhyaya, 2019; Xavier et al., 2019; Zettlemoyer et al., 2019), nine of which did not test for a relationship between tracking and the other studied traits (Suzuki & Kudo, 1997; Adrian et al., 2006; Xu et al., 2009; Szilvia et al., 2012; Bell et al., 2015; McDermott & DeGroote, 2016; Sherwood et al., 2017; Sharma & Upadhyaya, 2019; Xavier et al., 2019). We present data from the remaining papers in Tables S2-S3.

Most studies examined tracking as how a phenophase related to temperature (85% of all tracking metrics), followed by precipitation (10%, includes snow removal), followed by photoperiod (3%), the climate mode NAO (1%), water table depth (0.5%), and year (0.5%). Four of the 30 studies examined more than one major climate metric, though some measured many versions

of temperature and/or precipitation metrics (e.g., 15 precipitation and/or temperature metrics considered in Munson & Long, 2017).

2 Model

To give an example of how tracking and non-stationarity may be included in current models we modified a model from Chesson *et al.* (2004), which was originally conceptualized for annual plants with a seedbank. Although the model can be conceived of more generally, we use the language of annual plant germination for concreteness.

The model includes a suite of species traits, including (and particularly relevant for our aims) traits controlling species response to the environment via germination each year and traits related to how species may bet-hedge across years (via a seedbank), as well as traits relating to resource competition each year. Within-season dynamics are controlled by resource competition resulting in fitness differences, while interannual variation in the environment provides opportunities for coexistence via fluctuation-dependent mechanisms (i.e., niche differences resulting from different germination functions).

Across years, for a community of n species, the seedbank (N) of species i at time t+1 is determined by the survival (s) of seeds that did not germinate in last season $(1-g_i(t))$ plus new biomass (B_i) produced during the length of the growing season (δ) converted to seeds at rate ϕ :

$$N_i(t+1) = sN_i(t)(1 - g_i(t)) + \phi B_i(t+\delta)$$
(1)

The production of new biomass each season follows a basic R* competition model: new biomass production depends on its resource uptake (f(R) converted into biomass at rate c_i) less maintenance costs (m), with uptake controlled by a, u, and θ :

$$\frac{\mathrm{d}B_i}{\mathrm{d}t} = c_i f(R) - mB_i \tag{2}$$

$$f(R) = \frac{aR^{\theta}}{1 + auR^{\theta}} \tag{3}$$

With the initial condition:

$$B_i(t+0) = N_i(t)g_i(t)b_0 \tag{4}$$

where b_0 is the initial biomass per seed.

The resource (R) itself declines across a growing season due to uptake by all species and abiotic loss (ϵ) :

$$\frac{\mathrm{d}R}{\mathrm{d}t} = -\sum_{i=1}^{n} f_i(R)B_i - \epsilon R \tag{5}$$

Germination depends both on the traits of the species and on the environment that year. The fraction of seeds germinating for a species each year is determined by the distance between τ_i , a species characteristic, and $\tau_p(t)$, an attribute of the environment, which varies year-to-year. Germination fraction declines according to a Gaussian as the distance between τ_i and $\tau_p(t)$ grows (we refer to this distribution as the 'germination curve').

$$g_i(t) = g_{max}e^{-h(\tau_p(t) - \tau_i)^2}$$
 (6)

The model is designed for multiple conceptualizations (Chesson *et al.*, 2004); given our focus here, we consider $\tau_p(t)$ to represent the environmental (abiotic) start of the growing season that varies from year-to-year and refer to it as the 'environmental start time.' τ_i represents the 'intrinsic biological start time' for species *i*. How well matched a species is to its environment each year can be measured as τ_i - $\tau_p(t)$, or the distance between the intrinsic (biological) start time and the environmental start time.

Adding phenological tracking to model:

We adjust the biological start time, τ_i so that it can respond to the environment dynamically through what we refer to as tracking. Tracking (α , which can vary between 0 to 1) decreases the distance between τ_i and $\tau_p(t)$, i.e., moving the intrinsic start time closer to the environmental start time in that year, resulting in a higher germination fraction (e.g., species B in Fig. S2b-c).

$$\alpha_i \in [0, 1]$$

$$\hat{\tau}_i = \alpha_i \tau_p(t) + (1 - \alpha_i) \tau_i \tag{7}$$

Thus,

when
$$\alpha_i = 0, \hat{\tau}_i = \tau_i$$

when $\alpha_i = 1, \hat{\tau}_i = \tau_p(t)$

Simulations:

Using this model framework, we simulated a suite of two-species communities in stationary and non-stationary environments and examined persistence. As our interest is primarily in the role of environmental tracking, we focus on situations where species vary in their match to the environment through both the intrinsic biological start time (τ_i) combined with tracking (α), examining simulations where all species had some level of tracking. We also varied species' resource use efficiency (via c_i), yielding species with different R^* (where a lower R^* means a species can draw the resource down to a lower level and is thus considered the superior competitor). Each simulation was composed of two sequential parts: first, a 500-year stationary period where the underlying distribution of the environment does not change (but is stochastic, yielding year-to-year variation) followed by a 500-year non-stationary period where the underlying distribution of the environment shifts to an earlier start of season (see Fig. 3 in main text). Thus, only species that persisted through the first stationary period continued into the non-stationary period.

To simulate competitive dynamics, both species were initialized with a census size of N(0) = 100 per unit area, and the temporally varying parameters $R_0(t)$ and $\tau_p(t)$ were generated for the stationary and nonstationary periods. Species-specific parameters $(c_i, \tau_i, \text{ and } \alpha_i)$ were drawn from uniform random distributions with the ranges given in Table S1. All other parameters were identical between species (Table S1). Within-year R^* competition dynamics were solved using an ode solver (ode in the R package deSolve) and ended when the resource was drawn down to $min(R^*)$, i.e., the R^* value of the better resource competitor. The end of season biomass of each

species was converted to seeds, and the populations were censused. At each census, a minimum cutoff was applied to define extinction from the model. Note that "coexistence" in this model is defined by joint persistence through time and not by low density growth rate.

3 References

- Adrian, R., Wilhelm, S. & Gerten, D. (2006). Life-history traits of lake plankton species may govern their phenological response to climate warming. *Global Change Biology*, 12, 652–661.
- Arfin Khan, M.A.S., Beierkuhnlein, C., Kreyling, J., Backhaus, S., Varga, S. & Jentsch, A. (2018). Phenological sensitivity of early and late flowering species under seasonal warming and altered precipitation in a seminatural temperate grassland ecosystem. *Ecosystems*, 21, 1306–1320.
- Bell, J.R., Alderson, L., Izera, D., Kruger, T., Parker, S., Pickup, J., Shortall, C.R., Taylor, M.S., Verrier, P. & Harrington, R. (2015). Long-term phenological trends, species accumulation rates, aphid traits and climate: five decades of change in migrating aphids. *Journal of Animal Ecology*, 84, 21–34.
- Bock, A., Sparks, T.H., Estrella, N., Jee, N., Casebow, A., Schunk, C., Leuchner, M. & Menzel, A. (2014). Changes in first flowering dates and flowering duration of 232 plant species on the island of guernsey. Glob Chang Biol, 20, 3508–19.
- Brooks, S.J., Self, A., Powney, G.D., Pearse, W.D., Penn, M. & Paterson, G.L.J. (2017). The influence of life history traits on the phenological response of british butterflies to climate variability since the late-19th century. *Ecography*, 40, 1152–1165.
- Chesson, P., Gebauer, R.L.E., Schwinning, S., Huntly, N., Wiegand, K., Ernest, M.S.K., Sher, A., Novoplansky, A. & Weltzin, J.F. (2004). Resource pulses, species interactions, and diversity maintenance in arid and semi-arid environments. *Oecologia*, 141, 236–253.
- Diamond, S.E., Frame, A.M., Martin, R.A. & Buckley, L.B. (2011). Species' traits predict phenological responses to climate change in butterflies. *Ecology*, 92, 1005–1012.
- Dorji, T., Totland, O., Moe, S.R., Hopping, K.A., Pan, J. & Klein, J.A. (2013). Plant functional traits mediate reproductive phenology and success in response to experimental warming and snow addition in Tibet. *Global Change Biology*, 19, 459–472.
- Du, Y.J., Chen, J.R., Willis, C.G., Zhou, Z.Q., Liu, T., Dai, W.J., Zhao, Y. & Ma, K.P. (2017). Phylogenetic conservatism and trait correlates of spring phenological responses to climate change in northeast china. *Ecology and Evolution*, 7, 6747–6757.
- Goodenough, A.E., Hart, A.G. & Stafford, R. (2010). Is adjustment of breeding phenology keeping pace with the need for change? linking observed response in woodland birds to changes in temperature and selection pressure. *Climatic Change*, 102, 687–697.

- Ishioka, R., Muller, O., Hiura, T. & Kudo, G. (2013). Responses of leafing phenology and photosynthesis to soil warming in forest-floor plants. *Acta Oecologica-International Journal of Ecology*, 51, 34–41.
- Jing, J., Li, K. & Liu, Z.G. (2016). Effects of varying temperature on leaf phenology and herbivory of dominant tree species in subtropical evergreen broad-leaves forest in eastern china. *Polish Journal of Ecology*, 64, 53–69.
- Kharouba, H.M., Paquette, S.R., Kerr, J.T. & Vellend, M. (2014). Predicting the sensitivity of butterfly phenology to temperature over the past century. *Global Change Biology*, 20, 504–514.
- Ladwig, L.M., Chandler, J.L., Guiden, P.W. & Henn, J.J. (2019). Extreme winter warm event causes exceptionally early bud break for many woody species. *Ecosphere*, 10.
- Lasky, J.R., Uriarte, M. & Muscarella, R. (2016). Synchrony, compensatory dynamics, and the functional trait basis of phenological diversity in a tropical dry forest tree community: effects of rainfall seasonality. *Environmental Research Letters*, 11.
- McDermott, M.E. & DeGroote, L.W. (2016). Long-term climate impacts on breeding bird phenology in pennsylvania, usa. *Global Change Biology*, 22, 3304–3319.
- Moussus, J.P., Clavel, J., Jiguet, F. & Julliard, R. (2011). Which are the phenologically flexible species? a case study with common passerine birds. *Oikos*, 120, 991–998.
- Munson, S.M. & Long, A.L. (2017). Climate drives shifts in grass reproductive phenology across the western USA. *New Phytologist*, 213, 1945–1955.
- Park, D.S., Breckheimer, I., Williams, A.C., Law, E., Ellison, A.M. & Davis, C.C. (2019). Herbarium specimens reveal substantial and unexpected variation in phenological sensitivity across the eastern united states. *Philosophical Transactions of the Royal Society B-Biological Sciences*, 374.
- Post, E. & Stenseth, N. (1999). Climatic variability, plant phenology, and northern ungulates. *Ecology*, 80, 1322–1339.
- Sharma, S. & Upadhyaya, H.D. (2019). Photoperiod response of annual wild cicer species and cultivated chickpea on phenology, growth, and yield traits. *Crop Science*, 59, 632–639.
- Sherwood, J.A., Debinski, D.M., Caragea, P.C. & Germino, M.J. (2017). Effects of experimentally reduced snowpack and passive warming on montane meadow plant phenology and floral resources. *Ecosphere*, 8.
- Suzuki, S. & Kudo, G. (1997). Short-term effects of simulated environmental change on phenology, leaf traits, and shoot growth of alpine plants on a temperate mountain, northern Japan. *Global Change Biology*, 3, 108–115.
- Szilvia, K., Fehervari, P., Krisztina, N., Andrea, H. & Tibor, C. (2012). Changes in migration phenology and biometrical traits of reed, marsh and sedge warblers. *Central European Journal of Biology*, 7, 115–125.

- Vegvari, Z., Juhasz, E., Toth, J.P., Barta, Z., Boldogh, S., Szabo, S. & Varga, Z. (2015). Life-history traits and climatic responsiveness in noctuid moths. *Oikos*, 124, 235–242.
- Xavier, R.D., Leite, M.B. & Matos, D.M.D. (2019). Phenological and reproductive traits and their response to environmental variation differ among native and invasive grasses in a neotropical savanna. *Biological Invasions*, 21, 2761–2779.
- Xia, J.Y. & Wan, S.Q. (2013). Independent effects of warming and nitrogen addition on plant phenology in the inner mongolian steppe. *Annals of Botany*, 111, 1207–1217.
- Xu, Z.F., Hu, T.X., Wang, K.Y., Zhang, Y.B. & Xian, J.R. (2009). Short-term responses of phenology, shoot growth and leaf traits of four alpine shrubs in a timberline ecotone to simulated global warming, eastern tibetan plateau, china. *Plant Species Biology*, 24, 27–34.
- Zettlemoyer, M.A., Schultheis, E.H. & Lau, J.A. (2019). Phenology in a warming world: differences between native and nonâĂŘnative plant species. *Ecology Letters*.
- Zhang, J.H., Yi, Q.F., Xing, F.W., Tang, C.Y., Wang, L., Ye, W., Ng, I., Chan, T.I., Chen, H.F. & Liu, D.M. (2018). Rapid shifts of peak flowering phenology in 12 species under the effects of extreme climate events in macao. *Scientific Reports*, 8.
- Zhu, J.T., Zhang, Y.J. & Wang, W.F. (2016). Interactions between warming and soil moisture increase overlap in reproductive phenology among species in an alpine meadow. *Biology Letters*, 12, 4.

4 Tables & figures

Table S1: Table of parameter values, definitions, and units.

Parameter	Value(s)	parameter values, definitions, and Definition	Unit Unit		
$\frac{1}{N_i}$	$init cond N_i(0) =$		seeds per unit area		
1 1 2	$100; \min(N_i(t)) =$	i	seeds per unit area		
	$100, min(10) = 10^{-4}$				
s	0.8	survival of species i	unitless		
B_i	cf. Eqn 2	biomass of species i	biomass		
R_0	$\sim log N(\mu, \sigma) mu =$	annually varying initial value			
100	$log(2), \sigma = 0.2$	of resource at the beginning of	resource		
	$log(z), \sigma = 0.2$	the growing season			
	\sim Unif(8,20)	conversion efficiency of R to	biomass resource		
c_i	\sim 0 IIII(8,20)		resource		
	0.05	biomass of species i	1		
m	0.05	maintenance costs during	$days^{-1}$		
	20	growth season i	1 -1		
a	20	uptake increase as R increases	$days^{-1}$		
		for species i	(days)(biomass)		
u	1	inverse of max uptake for	resource		
		species i			
θ	1	shape of uptake for species i	unitless		
ϕ	0.05	conversion of end-of-season	biomass ⁻¹ , but concep-		
		biomass to seeds	tually $\frac{\text{seeds}}{(\text{biomass})(\text{seeds})}$		
ϵ	1	abiotic loss of R	$days^{-1}$		
g_{max}	0.5	max germination rate of	unitless		
		species			
h	100	controls the the rate at which	$days^{-2}$		
		germination declines as τ_p			
		deviates from optimum for			
		species i			
g_i	cf. Eqn 6	germination rate	unitless		
$ au_p$	$\sim \beta(10, 10)$	timing of pulse	days		
$ au_i$	$\sim \text{Unif}(0.1,0.9)$	timing of max germination of	days		
		species i			
α_i	$\sim \text{Unif}(0,1)$	phenological tracking of	unitless		
		species i			
b_0	1	biomass of a seedling	biomass seeds		
f(R)	cf. Eqn 3	resource uptake rate for	resource		
• \ /	•	species i	(days)(biomass)		
t	1	annual timestep	years		
$0 \to \delta$	determined by rate	time during the growing sea-	days		
J / J	of resource depletion	son			

Table S2: Summary of traits related to phenological tracking in the literature and whether papers reported statistical evidence that they were linked or not. See Table S2 for an extended version.

Trait	n linked	linked n not linked	
diet traits	0	3	
early/late phenophase	10	4	
habitat traits	1	3	
height	1	0	
hibernation stage	0	3	
leaf/shoot size	1	0	
migration traits	3	2	
mobility	1	3	
nativeness	1	3	
niche breadth	3	2	
other Lepidopteran traits	3	4	
other bird traits	1	1	
other leaf traits	4	3	
other plant traits	1	1	
overwintering	2	1	
range traits	1	4	
root traits	3	0	
seed weight/size/number	1	2	
woody/herbaceous	1	0	

Table S3: Summary of results from literature on phenological tracking showing which phenophases researchers found were linked to which traits, or not.

Taxa	Phenophase	Trait	n linked	n not linked
Lepidoptera	activity length	hibernation stage		1
Lepidoptera	activity length	migration traits		1
Lepidoptera	activity length	other Lepidopteran traits	1	
Lepidoptera	appearance/collection date	diet traits		1
Lepidoptera	appearance/collection date	early/late phenophase	2	
Lepidoptera	appearance/collection date	habitat traits		2
Lepidoptera	appearance/collection date	hibernation stage		1
Lepidoptera	appearance/collection date	migration traits	1	
Lepidoptera	appearance/collection date	mobility		2
Lepidoptera	appearance/collection date	niche breadth	2	1
Lepidoptera	appearance/collection date	other Lepidopteran traits	1	2
Lepidoptera	appearance/collection date	overwintering	2	
Lepidoptera	appearance/collection date	range traits	1	2
Lepidoptera	flight timing	early/late phenophase	1	1
Lepidoptera	flight timing	mobility	1	1
Lepidoptera	flight timing	niche breadth		1
Lepidoptera	flight timing	other Lepidopteran traits		1
Lepidoptera	flight timing	overwintering		1
Lepidoptera	flight timing	range traits		1
Lepidoptera	last/median emergence dates	diet traits		1
Lepidoptera	last/median emergence dates	habitat traits		1
Lepidoptera	last/median emergence dates	hibernation stage		1
Lepidoptera	last/median emergence dates	migration traits		1
Lepidoptera	last/median emergence dates	other Lepidopteran traits	1	1
passerine birds	breeding time	diet traits		1
passerine birds	breeding time	habitat traits	1	
passerine birds	breeding time	migration traits	2	
passerine birds	breeding time	niche breadth	1	
passerine birds	breeding time	other bird traits	1	1
plants	budbreak/leafing	early/late phenophase	3	1
plants	budbreak/leafing	nativeness		1
plants	budbreak/leafing	other leaf traits	2	1
plants	budbreak/leafing	range traits		1
plants	flowering/fruiting	early/late phenophase	4	2
plants	flowering/fruiting	height	1	
plants	flowering/fruiting	leaf/shoot size	1	
plants	flowering/fruiting	nativeness	1	2
plants	flowering/fruiting	other leaf traits	$\frac{1}{2}$	2
plants	flowering/fruiting	other plant traits	1	1
plants	flowering/fruiting	root traits	3	-
plants	flowering/fruiting	seed weight/size/number	1	2
plants	flowering/fruiting	woody/herbaceous	1	_

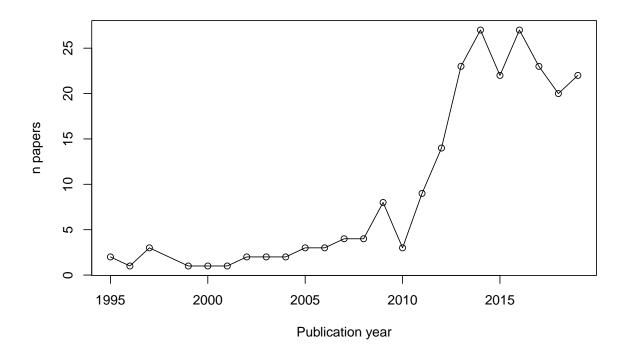
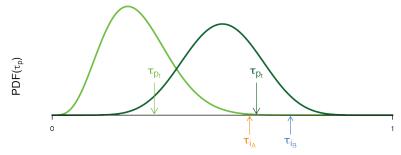
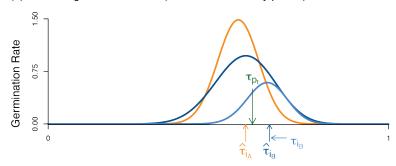


Figure S1: Trends in all papers using search terms over time. Of papers from which we could extract data all 25 of 30 were published in 2011 or onward.

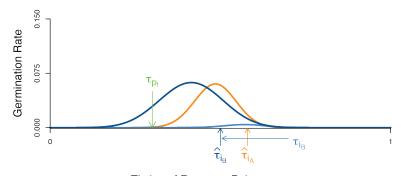
(a) Distributions of resource timing under stationary and non-stationary periods



(b) Realized germination rates (draw from stationary period)



(c) Realized germination rates (draw from end of non-stationary period)



Timing of Resource Pulse, τ_{p}

Figure S2: The distributions of the environment (a) and species' germination for two sample years (b-c) in our seed germination model. (a) The timing of the resource pulse $(\tau_p(t))$, which defines the environmental start of season, is β -distributed with parameters $\beta(10, 10)$ during the stationary period (dark green) shifting to $\beta(5, 15)$ through the nonstationary period (light green). (b) Realized germination rate as a function of $\tau_p(t)$ for two species during the stationary period: the orange line is a non-tracking species A with preferred germination time, τ_{iA} , that is close to the mean of the stationary period; the blue lines show the difference in realized germination rate of a tracking species with a preferred germination time, τ_{iB} , that is further from the mean of the stationary period both without (light blue) and with (dark blue) the effect of tracking; note the shift from τ_{iB} to τ_{iB} . (c) Realized germination rate of species A and species B at the end of the nonstationary period. Note the change in axes from (b) to (c) shows the decline in overall germination rate as the environment moves away from the preferred germination time of both species.

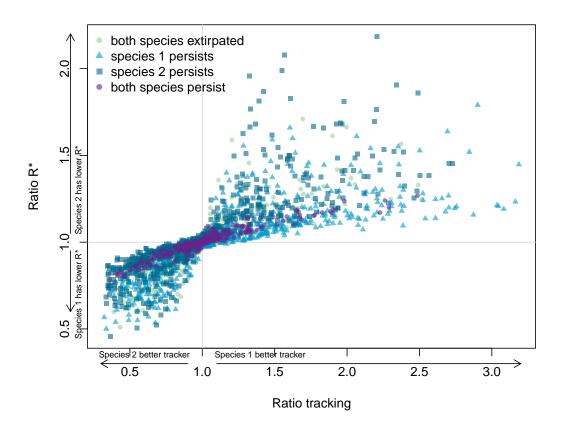


Figure S3: How non-stationarity reshapes two-species communities in a simple model where tracking (X axis: species 1/species 2) trades off with R^* (Y axis: species 1/species 2): each point represents one two-species community that persisted through 500 years of stationary dynamics while the shape and color represent the outcome for that two-species community of 500 years of non-stationarity, where the abiotic start of the season shifts earlier.