Supplementary Information

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Appendix S1. Methods for reconstruction of DBH

This is still rough/ mostly notes.

For each core, DBH can be reconstructed outside-in (based on recent DBH, subtracting growth recorded in tree rings) or inside-out (summing RW from the inside out). We generally gave precedence to the outside-in approach. Specifically, when DBH was taken at the time of coring,

At some of our sites where DBH was not taken at the time of coring (SCBI,), DBH measurements taken before or slightly after the time of coring could be used. (see issue #19 in ForestGEO_dendro) If before, ... If after... For all outside-in reconstructions, if a negative DBH was predicted...

When there were more than one cores for a tree, the DBH reconstructions from each core were averaged to produce a single estimate of the tree's DBH through time. When the start or end dates of the records from the cores differed, we extrapolated growth of the shorter core to match the years covered by the longer core. Specifically, to fill in years at the more recent end, we assumed that the average growth rate of the ten years prior to the missing records applied to the missing years. To fill in years at the beginning of the tree's lifespan, we likewise assumed that the ten years adjacent to the missing record applied to the missing years; however, if this yielded a negative DBH estimate for the earliest year in the reconstruction, we divided the existing minimum DBH by number of years missing and applied that value to each year. We note that these reconstructed growth records were used only for the reconstruction of DBH and were not included as response variables in any of our analyses.

In either case we need bark thickness—ideally allometries describing the relationship between DBH and bark thickness (Table S4). This is especially critical for thick-barked species. When bark thickness data were available, we generated allometries (issue #8 in ForestGEO_dendro)... lognormal model with intercept forced to zero: lm(bark_depth.mm ~ -1 + log(dbh_no_bark.cm+1):bark_species, data = bark). When bark thickness data were not available, we used published bark allometries from other sources (Table S4)

Appendix S2. Methods for comparing climwin results with traditional methods

(ISSUE #35 in ForestGEO-climate-sensitivity)

To verify that our methods gave similar results to traditional methods, we conducted qualitative comparisons of our results to previous studies based on the same cores (Table S5). We also conducted a formal comparison using identical tree-ring and climate data for four well-studied species: PSME (Cedar Breaks, Utah), ABAL (Zofin), PIMA (Scotty Creek), and LITU (SCBI) (Fig. S1). We compared results from an analysis using conventional methods, as detailed below, to an analysis using our method as described in the Methods section, but with the *climwin* climate variable selection process limited to just the species of interest (as opposed to all species at the site), climate variables considered individually rather than additively, and with start date adjusted to match the conventional method.

The ring-width series from each core was standardized via ARSTAN using a 2/3rds n spline, where n is the number of years in the series (Cook, 1985; Cook & Kairiukstis, 1990- citations in Helcoski). (The following italic text is plagarized from Helcoski and needs to be reworded:) The influence of outliers in all series was reduced using the adaptive power transformation, which also stabilises the variance over time (Cook & Peters, 1997). Next, each series was stabilised using either the average correlation between raw ring-width series (rbar) method or a 1/3rds spline method to adjust changes in variance as series replication decreased towards the earlier portion of each chronology (Jones et al., 1997). The 1/3rds spline method was chosen when replication in the inner portion of each chronology (c. the inner 30-50 yr of each record depending on full chronology length) dropped below three trees. Once that step was complete, a robust biweight mean chronology for each species was calculated from the ring-width indices (Cook, 1985). We chose to use residual chronologies because the autoregressive standardisation process in creating them removes much of the tree-level autocorrelation in growth and these chronologies would most likely contain the most conservative information on drivers of interannual growth (Cook, 1985).

Following Helcoski et al. (2019), we defined chronology start dates according to the subsample signal strength (SSS), using a cutoff of SSS = 0.80 (or 80% of the population signal). Thus, for this analysis only, we defined chronology start dates as the year the SSS exceeded 0.80 or two years after the start of the climate record, whichever came later. SSS exceeded 0.80 well before the start of the 1901 start of climate records for PSME (1800s), ABAL (1700), and PIMA (1850s). For LITU, SSS reached 0.8 with 11 trees in 1919, which we used as the start date for this series. We note that these start date criteria differ from those used in the main analysis (Table S3), which had earlier start dates because the analysis was not constrained by a need to represent the full population signal. End dates were defined as the last full year prior to sampling (Table S3).

Appendix S3. Dealing with rapidly changing climate and tree growth

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Our analysis included two sites where climate change has had pronounced effects on tree growth: Scotty Creek, NW Territories, Canada (SC) and Little Tesuque, New Mexico, USA (LT). At SC, [temperatures have increased by X $^{\circ}$ over X years]..., resulting in negative growth trends in basal area index (BAI) starting around 1950 and significant growth declines since 1970 in 56% of trees (Sniderhan & Baltzer, 2016). At LT, (drought has increased dramatically), resulting in many missing rings in recent years.

This is in process. We will try and compare 3 methods: (1) our standard approach, (2) detrending the climate variables (#53), (3) applying the climwin step only for older records-before the most rapid climate change. We will work with SC and LT researchers to determine which makes most sense, and use that as the main approach for these sites.

Table S1. Site Details

site code	site name	latitude	longitude	elevation (m.a.s.l.)	cores within ForestGEO plot?	canopy positions	tree statuses	date range	dormant season	months in climwin
BCI	Barro Colorado Island	9.15430	-79.8461	120-160	no	canopy	live, dead	1931-2014	Nov-Apr	pOct-cDec
HKK	Huai Kha Khaeng	15.63240	99.2170	549-638	no	all	live	1903-2011	Nov-Apr	pOct-cDec
LT	Little Tesuque	35.73838	-105.8382	2682	n.a.	all	live	1903-2018	•	pMay-cAug
CB	Utah Forest Dynamics Plot	37.66150	-112.8525	3020-3169	yes		live	1903-2007		pMay-cAug
SCBI	Smithsonian Conservation Biology Institute	38.89350	-78.1454	273-338	yes	all	live, dead	1903-2017	Oct-Apr	pMay-cAug
LDW	Lilly Dickey Woods	39.23590	-86.2181	230-303		canopy	live, dead	1903-2019		pMay-cAug
HF	Harvard Forest	42.53880	-72.1755	340-368	ves	all	live, dead	1903-2014		pMay-cAug
NE	Niobrara/Halsey	42.78000	-100.0210	644-702	some	canopy	live		Oct-Apr	pMay-cAug
ZOF	Žofín Forest Dynamics Plot	48.66380	14.7073	736-829	some	all	live, dead	1903-2013	Oct-Mar	pMay-cAug
SC	Scotty Creek	61.30000	-121.3000	258-274	no	all	live, dead	1903-2013		pMay-cAug

Table S2. Species analyzed, their characteristics, and bark allometries applied

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NOTE: bark.allometry field is not yet right—we will have just one latin name per site, corresponding to allometries in Table S4. But it does give correct info for what is currently applied. We also intend to find and apply more allometries.

species code	family	latin name	sites sampled	leaf type	leaf phenology	light requirements	bark allometry
ABAL	Pinaceae	Abies alba	ZOF	needleleaf	evergreen	shade-tolerant	neglected in Zofin
ABBI	Pinaceae	Abies bifolia	CB	needleleaf	evergreen		neglected in CedarBreaks
ACRU	Sapindaceae	Acer rubrum	HF	broadle af	deciduous (cold)		acru in HarvardForest
ACSA	Sapindaceae	Acer saccharum	LDW	broadleaf	deciduous (cold)		acru in LillyDickey, acru in LillyDickey
AFXY	Fabaceae	Afzelia xylocarpa	HKK	broadle af	deciduous (drought)		neglected in HKK
BEAL	Betulaceae	Betula alleghaniensis	HF	broadleaf	deciduous (cold)		Betula alleghaniensis in HarvardForest
BEPA	Betulaceae	Betula papyrifera	NE	broadle af	deciduous (cold)		Betula papyrifera in Nebraska
CACO	Juglandaceae	Carya cordiformis	SCBI	broadle af	deciduous (cold)		caco in SCBI
CAGL	Juglandaceae	Carya glabra	SCBI	broadle af	deciduous (cold)		cagl in SCBI
CAOV	Juglandaceae	Carya ovata	LDW	broadle af	deciduous (cold)		cagl in LillyDickey
CAOVL	Juglandaceae	Carya ovalis	SCBI	broadle af	deciduous (cold)		caovl in SCBI
CATO	Juglandaceae	Carya tomentosa	SCBI	broadle af	deciduous (cold)		cato in SCBI
CHTA	Meliaceae	Chukrasia tabularis	HKK	broadleaf	brevi-deciduous (drought)		neglected in HKK
FAGR	Fagaceae	Fagus grandifolia	HF, SCBI	broadleaf	deciduous (cold)		neglected in HarvardForest, neglected in LillyDickey, neglected in SCBI
FASY	Fagaceae	Fagus sylvatica	ZOF	broadle af	deciduous (cold)	shade-tolerant	neglected in Zofin
FRAM	Oleaceae	Fraxinus americana	LDW, SCBI	broadleaf	deciduous (cold)		Fraxinus ssp. in Lilly Dickey, fram in SCBI
FRNI	Oleaceae	Fraxinus nigra	SCBI	broadle af	deciduous (cold)		fram in SCBI
JACO	Bignoniaceae	Jacaranada copaia	BCI	broadle af	deciduous (drought)	light-demanding	JCO in BCI
JUNI	Juglandaceae	Juglans nigra	SCBI	broadle af	deciduous (cold)		juni in SCBI
JUVI	Cupressaceae	Juniperus virginiana	NE				neglected in Nebraska
LITU	Magnoliaceae	Liriodendron tulipifera	LDW, SCBI	broadleaf	deciduous (cold)		litu in LillyDickey, litu in LillyDickey, litu in SCBI
MEAZ	Meliaceae	Melia azedarach	HKK	broadle af	deciduous (drought)	light-demanding	neglected in HKK
PIAB	Pinaceae	Picea abies	HF	needleleaf	evergreen	intermediate	neglected in HarvardForest, neglected in Zofin
PIEN	Pinaceae	Picea engelmannii	СВ	needleleaf	evergreen		Picea engelmannii in CedarBreaks
PIFL	Pinaceae	Pinus flexilis	CB	${\it needle leaf}$	evergreen		Pinus monticola in CedarBreaks
PILO	Pinaceae	Pinus longaeva	CB	${\it needle leaf}$	evergreen		neglected in CedarBreaks
PIMA	Pinaceae	Picea mariana	SC	${\it needle leaf}$	evergreen		PIMA in ScottyCreek
PIPO	Pinaceae	Pinus ponderosa	NE, LT	needleleaf	evergreen		Pinus jeffreyi in Little Tesuque, Pinus jeffreyi in Nebraska
PIPU	Pinaceae	Picea pungens	CB	${\it needle leaf}$	evergreen		neglected in CedarBreaks
PIST	Pinaceae	Pinus strobus	HF, SCBI	needleleaf	evergreen		Pinus strobus in HarvardForest, pist in SCBI
PIST2	Pinaceae	Pinus strobiformis	LT	needleleaf	evergreen		Pinus monticola in Little Tesuque
POTR	Salicaceae	Populus tremuloides	СВ	broadleaf	deciduous (cold)		Populus tremuloides in CedarBreaks
PSME	Pinaceae	Pseudotsuga menziesii	СВ	needleleaf	evergreen		Pseudotsuga menziesii in CedarBreaks
QUAL	Fagaceae	Quercus alba	LDW, SCBI	broadle af	deciduous (cold)		qual in Lilly Dickey, qual in SCBI
QUMO	Fagaceae	Quercus montana	LDW, SCBI	broadleaf	deciduous (cold)		qupr in LillyDickey, qupr in SCBI
QURU	Fagaceae	Quercus rubra	HF, LDW, SCBI	broadleaf	deciduous (cold)		quru in HarvardForest, Quercus rubra in LillyDickey, quru in SCBI
QUVE	Fagaceae	Quercus velutina	LDW, SCBI	broadleaf	deciduous (cold)		quve in Lilly Dickey, quve in SCBI
TEPA	Burseraceae	Tetragastris panamensis	BCI	broadle af	evergreen	shade-tolerant	TPA in BCI
TOCI	Meliaceae	Toona ciliata	HKK	broadle af	deciduous (drought)		neglected in HKK
TRTU	Meliaceae	Trichilia tuberculata	BCI	broadle af	evergreen	shade-tolerant	TTU in BCI
TSCA	Pinaceae	Tsuga canadensis	HF	needleleaf	evergreen		Tsuga canadensis in HarvardForest

*Bark allometry field indicates the species and site sampled to construct the bark allometry. When neither raw data nor an allometric equation for the study species was available, we selected the most appropriate equation that could be located for similar species. Equations are given in Table S4.

Table S3. Sampling details for species by site

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site	species code	n trees all	n cores all	n trees dbh	n cores dbh	dbh range sampled	dbh range reconstructed*	date range
BCI	JACO	12	18	11	17	30.2-63.5	2.6-56.4	1931-2014
BCI	TEPA	18	29	17	26	22.1-59.5	2.7-49.4	1931-2014
BCI	TRTU	23	37	20	31	20.7-43.6	4.8-41.5	1931-2014
$^{\mathrm{CB}}$	ABBI	22	41	20	37	13.9-54.2	0.9-46.4	1903-2000
$^{\mathrm{CB}}$	PIEN	12	21	10	15	14-54.9	0.9-33.1	1903-2000
$^{\mathrm{CB}}$	PIFL	13	22	12	21	17.6-64.1	1.5-47.5	1903-1998
$^{\mathrm{CB}}$	PILO	11	16	8	12	45.8-121.6	35.4-118.5	1903-1999
$^{\mathrm{CB}}$	PIPU	15	28	15	28	22.4-50.8	9.5-48.4	1903-2000
$^{\mathrm{CB}}$	POTR	17	27	17	26	23.6-47.6	4.3-35.4	1903-2000
$^{\mathrm{CB}}$	PSME	11	20	10	18	20.7-64.2	0.5-41.5	1903-1999
$_{ m HF}$	ACRU	18	59	18	59	10.1-22.1	0.9-20.4	1903-2013
$_{ m HF}$	BEAL	13	44	13	44	10.2-37.9	0-17.2	1904-2013
$_{ m HF}$	QURU	74	180	73	177	19.5-53	1.1-48.3	1903-2014
$_{ m HF}$	TSCA	32	83	32	71	10.6-37	0-28.4	1923-2014
HKK	AFXY	39	127	39	127	20.1-98.7	0.1-81.4	1903-2011
HKK	CHTA	28	70	28	70	16-64.6	0.2-59.5	1904-2010
HKK	MEAZ	46	130	46	130	25.6-98.1	3.8-80.3	1914-2011
HKK	TOCI	45	143	45	143	16.6-116.4	1.7-80.5	1903-2011
LDW	ACSA	35	66	34	64	9-64.6	0-52.4	1903-2019
LDW	CAOV	9	18	8	15	NA-NA	1.4-37.4	1903-2013
$_{ m LDW}$	LITU	15	28	14	25	NA-NA	1.2-69.4	1903-2019
LDW	QUAL	10	20	NA	NA	NA	NA	1903-2013
LDW	QUMO	10	20	8	16	NA-NA	1.1-52.4	1903-2013
LDW	QUVE	9	18	NA	NA	NA	NA	1903-2013
NE	BEPA	28	84	28	84	NA-NA	0-25.5	1903 - 1995
NE	JUVI	29	60	29	60	16.6-21.5	0.3-18.4	1938-1994
NE	PIPO	71	134	70	129	19.3-64	0-41.1	1934-1994
LT	PIPO	10	20	10	20	23.2-52.8	14.4-48.4	1903-2018
LT	PIST2	7	14	7	14	25.7-39.8	3.9-34.5	1903-2018
SCBI	CACO	15	15	14	14	10.62-38.52	1.7-32.2	1903-2015
SCBI	CAGL	39	39	36	36	10.28-52.31	1.6-49.3	1903-2015
SCBI	CAOVL	25	25	24	24	15.11-60.32	2.6-47.2	1903-2015
SCBI	CATO	15	15	14	14	12.86-35.95	3.7-28.4	1903-2015
SCBI	FAGR	76	76	74	74	10.05-41.02	0.1-41.2	1920-2009
SCBI	FRAM	66	66	61	61	8.11-94.73	0.1-84.4	1903-2016
SCBI	FRNI	12	12	12	12	11.04-39.2	0.5-27.3	1903-1996
SCBI	JUNI	30	30	28	28	20.4-76.19	5.6-59.5	1903-2010
SCBI	LITU	106	106	104	104	10-91.42	0.1-81.1	1903-2010
SCBI	PIST	36	36	36	36	13.92-50.96	1.6-45.2	1931-2010
SCBI	QUAL	66	66	66	66	11.4-76.73	0.3-70.4	1903-2009
SCBI	QUMO	67	67	67	67	10.22-84.59	0.3-69.5	1903-2017
SCBI	QURU	70	70	70	70	11.07-87.65	2.5-79.2	1903-2016
SCBI	QUVE	81	81	81	81	16.02-82.33	0.5-78.4	1903-2009
SC	PIMA	442	442	101	101	7-14.9	0.5-12.5	1903-2013
ZOF	ABAL	55	55	50	50	41-121	21.3-108.5	1903-2011
ZOF	FASY	1369	1369	1358	1358	NA-NA	0.1-115.3	1903-2013
ZOF	PCAB	644	644	599	599	NA-NA	0.5 - 126.4	1903-2011

^{*}Maximum reconstructed DBH's analyzed are less than maximum sampled DBH's because we discard size ranges with < 3 conspecific trees.

Table S4. Allometric equations for bark thickness

species	equation	n	DBH.range.cm	site	source
Acer rubrum	bark.mm = 0.619*log(dbh.cm + 1)	10	8.2-39.6	SCBI	Anderson-Teixeira et al. (2015)
Betula alleghaniensis	bark.mm = (0.15 + 0.03*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Betula papyrifera	bark.mm = (0.13 + 0.05*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Carya cordiformis	bark.mm = 0.793*log(dbh.cm + 1)	9	5.9-68.2	SCBI	Anderson-Teixeira et al. (2015)
Carya ovalis	bark.mm = 1.531*log(dbh.cm + 1)	8	6.4-63.1	SCBI	Anderson-Teixeira et al. (2015)
Carya ovata	bark.mm = 1.035*log(dbh.cm + 1)	8	19.1-78	SCBI	Anderson-Teixeira et al. (2015)
Carya tomentosa	bark.mm = 1.105*log(dbh.cm + 1)	8	5-57.3	SCBI	Anderson-Teixeira et al. (2015)
Fraxinus americana	bark.mm = 2.223*log(dbh.cm+1)	9	6.1-94.2	SCBI	Anderson-Teixeira et al. (2015)
Fraxinus americana	bark.mm = (0.38 + 0.05*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Jacaranada copaia	bark.mm = 2.993*log(dbh.cm + 1)	5	45.6-75	Panama	Raquel Alfaro-Sánchez (unpublished data)
Juglans nigra	bark.mm = 2.107*log(dbh.cm + 1)	9	13.6-85.4	SCBI	Anderson-Teixeira et al. (2015)
Liriodendron tulipifera	bark.mm = 1.637*log(dbh.cm + 1)	9	27.5-136.5	SCBI	Anderson-Teixeira et al. (2015)
Picea engelmannii	bark.mm = (0.15 + 0.04*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Picea mariana	bark.mm = 3.726*log(dbh.cm + 1)	12	6.9-7.9	Scotty Creek	Anastasia Sniderhan and Jennifer Baltzer (unpublished data)
Pinus flexilis	$bark.mm = (1.299 * \sqrt{(dbh.cm)^{0.609}})^2$	29	10-130	California (3 montane sites)	Zeibig-Kichas et al. (2016)
Pinus ponderosa	$bark.mm = (1.298 * \sqrt{(dbh.cm)^{0.802}})^2$	81	5-160	California (4 montane sites)	Zeibig-Kichas et al. (2016)
Pinus strobus	bark.mm = 1.568 * log(dbh.cm + 1)	1	28.4-28.4	Illinois	Miles and Smith (2009)
Pinus strobus	bark.mm = (0.02 + 0.10*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Pinus strobus	bark.mm = (0.02 + 0.10*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Populus tremuloides	bark.mm = (0.10 + 0.07*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Pseudotsuga menziesii	$bark.mm = (0.785*\sqrt(dbh.cm))^2$	30	10-200	California (3 montane sites)	Zeibig-Kichas et al. (2016)
Pseudotsuga menziesii	bark.mm = (0.40 + 0.17*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Quercus alba	bark.mm = 1.828*log(dbh.cm + 1)	10	9.3-101.8	SCBI	Anderson-Teixeira et al. (2015)
Quercus montana	bark.mm = 2.083*log(dbh.cm + 1)	8	5.8-99.1	SCBI	Anderson-Teixeira et al. (2015)
Quercus rubra	bark.mm = 0.98*log(dbh.cm + 1)	10	24.1-143.2	SCBI	Anderson-Teixeira et al. (2015)
Quercus rubra	bark.mm = (0.19 + 0.07*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.
Quercus velutina	bark.mm = 1.394*log(dbh.cm + 1)	8	16.2-110.7	SCBI	Anderson-Teixeira et al. (2015)
Tetragastris panamensis	bark.mm = 1.672*log(dbh.cm + 1)	4	22.7-48.8	Panama	Raquel Alfaro-Sánchez (unpublished data)
Trichilia tuberculata	bark.mm = 1.367*log(dbh.cm + 1)	12	21-40.5	Panama	Raquel Alfaro-Sánchez (unpublished data), Pete Kerby-Miller and Helene Muller-Landau (unpublished data)
Tsuga canadensis	bark.mm = (0.18 + 0.08*dbh.cm)/2	NA		North America average of its range	Miles, Patrick D.; Smith, W. Brad. 2009.

For assignments of species as proxies for those with out available bark allometries, see Table S2.

Table S5. Qualtiative comparison of results from this study with previous studies employing conventional methods $\,$

	Precipitati	on response	Temperatu	re response		
species	previously observed	observed here	previously observed	observed here	reference	
Barro Co	lorado Island, Panan	na				
JACO	pos. correlation to Apr-Dec <i>PPT</i> (strongest of the 3 species)	pos. correlation to Mar-Dec <i>PPT</i> (strongest of the 3 species)	no sig. correlation to annual T_{mean} or T_{min}	neg. response to Feb-Mar T_{\min}	Alfaro-Sánchez et al. 2017	
TEPA	pos. correlation to Apr-Dec <i>PPT</i> (response weaker than JACO, similar to TRTU)	pos. correlation to Mar-Dec <i>PPT</i> (response weaker than JACO, similar to TRTU)	no sig. correlation to annual T_{mean} or T_{min}	no sig. correlation to Feb-Mar T_{min}	Alfaro-Sánchez et al. 2017	
TRTU	pos. correlation to Apr-Dec <i>PPT</i> (response weaker than JACO, similar to TEPA)	pos. correlation to Mar-Dec PPT(response weaker than JACO, similar to TEPA)	no sig. correlation to annual T_{mean} or T_{min}	non-sig. slight pos. response to Feb-Mar T_{min}	Alfaro-Sánchez et al. 2017	
	a Khaeng, Thailand					
AFXY	sig. pos. correlation with June <i>PPT</i> , otherwise n.s.	slight concave-down response to p.Sept-June <i>PPT</i> frequency	sig. neg. correlation with T_{max} in Aug and Dec; T_{min} in p.Oct., Jul, Aug	slight concave-down response to Apr-Oct T_{max}	Vlam et al. 2013	
CHTA	sig. pos. correlation with April PPT , otherwise n.s.	slight concave-down response to p.Sept-June PPT frequency	sig. neg. correlation with T_{max} in May, Aug-Sept; T_{min} in Feb, May, Aug	slight neg. response to Apr-Oct T_{max}	Vlam et al. 2013	
MEAZ	sig. pos. correlation with April PPT , otherwise n.s.	concave-down response to p.Sept-June PPT frequency	sig. neg. correlation with T_{max} in May-Aug; T_{min} in May-Aug	neg. response to Apr-Oct T_{max}	Vlam et al. 2013	
TOCI	sig. pos. correlation with p.Oct-p.Nov and April-May PPT	concave-down /increasing response to p.Sept-June PPT frequency	sig. neg. correlation with T_{max} every month from pOct-June (excluding March); T_{min} in Jan and Mar-Aug	neg. response to Apr-Oct T_{max}	Vlam et al. 2013	
PIPO			1148		-	
PIST2					-	

Table S5, cont.

	Precipitati	on response	Temperatu	re response		
species	previously observed	observed here	previously observed	observed here	reference	
Smithsonia	an Conservation Biol	ogy Institute, Virg	inia, USA			
CACO	pos. correlations with May-Aug <i>PPT</i> (sig. May, July)	NA	neg. correlations with May-Aug <i>PET</i> (sig. May-July)	**(sig or ns)** neg. response to May-July PET	Helcoski et al. 2019	
CAGL	pos. correlations with May-Aug PPT (sig. May)	NA	neg. correlations with May-Aug PET (n.s.)	**(sig or ns)** neg. response to May-July PET	Helcoski et al. 2019	
CAOVL	pos. correlations with May-Aug PPT (sig. Aug)	NA	neg. correlations with May-Aug PET (sig. all months)	**(sig or ns)** neg. response to May-July PET	Helcoski et al. 2019	
CATO	pos. correlations with May-Aug PPT (n.s.)	NA	neg. correlations with May-Aug PET (sig. June)	**(sig or ns)** neg. response to May-July PET	Helcoski et al. 2019	
FAGR	pos. correlations with May-Aug PPT (sig. July-Aug)	NA	neg. correlations with May-Aug PET (sig. July-Aug)	**(sig or ns)** neg. response to May-July PET	Helcoski et al. 2019	
FRAM	pos. correlations with May-Aug PPT (sig. May-June)	NA	neg. correlations with May-Aug PET (sig. May-June)	**(sig or ns)** neg. response to May-July PET	Helcoski et al 2019	
FRNI	no sig. correlations with peak growing season PPT	NA	no sig. correlations with peak growing season PET	**(sig or ns)** neg. response to May-July PET	Helcoski et al 2019	
JUNI	pos. correlations with May-Aug PPT (sig. Jun-Aug)	NA	neg. correlations with May-Aug PET (sig. July-Aug)	non-sig. neg. response to May-July PET	Helcoski et al 2019	
LITU	pos. correlations with May-Aug PPT (sig. May-July)	NA	neg. correlations with May-Aug PET (sig. all months)	**(sig or ns)** neg. response to May-July PET	Helcoski et al 2019	
PIST	pos. correlations with May-Aug PPT (n.s.)	NA	neg. correlations with May-Aug PET (n.s.)	**(sig or ns)** neg. response to May-July PET	Helcoski et al 2019	
QUAL	pos. correlations with May-Aug PPT (sig. May)	NA	neg. correlations with May-Aug PET (sig. all months)	**(sig or ns)** neg. response to May-July PET	Helcoski et al 2019	
QUMO	pos. correlations with May-Aug <i>PPT</i> (sig. May)	NA	(sig. An Holitis) neg. correlations with May-Aug PET (sig. May-June, Aug)	**(sig or ns)** neg. response to May-July PET	Helcoski et al 2019	
QURU	pos. correlations with May-Aug PPT (n.s.)	NA	neg. correlations with May-Aug PET (sig. May, July-Aug)	**(sig or ns)** neg. response to May-July PET	Helcoski et al. 2019	
QUVE	pos. correlations with May-Aug <i>PPT</i> (sig. May-July)	NA	neg. correlations with May-Aug PET (sig. all months)	**(sig or ns)** neg. response to May-July PET	Helcoski et al. 2019	

Table S5, cont.

	Precipitati	on response	Temperatu	re response	
species	previously observed	observed here	previously observed	observed here	reference
Lilly Dick	ey Woods, Indiana,	USA			
LITU	pos. correlations with Jun-Aug PDSI	pos. response to June PPT	neg. response to Jun-Aug T_{\max}	neg. response to June PET	Maxwell, Harley, and Robeson 2016
QUAL	pos. correlations with Jun-Aug PDSI	pos. response to June PPT	neg. response to Jun-Aug T_{max}	neg. response to June PET	Maxwell, Harley, and Robeson 2016
QUMO	pos. correlations with Jun-Aug PDSI	pos. response to June PPT	neg. response to Jun-Aug T_{max}	neg. response to June PET	Maxwell, Harley, and Robeson 2016
QUVE	pos. correlations with Jun-Aug PDSI	pos. response to June PPT	neg. response to Jun-Aug T_{max}	neg. response to June PET	Maxwell, Harley, and Robeson 2016
Harvard I	Forest, Massachusetts	s, USA			
Nichagas	and Handley Nahna	le TICA			-
Niobrara BEPA	and Hansley, Nebras little relationship to ppt within analysis timeframe (exception: pos. corr. with pAug pre); stronger relationship to streamflow and	ka, USA	little relationship to T_{mean} within analysis timeframe (exception: neg. corr. with pJune and cJan T_{mean})		Bumann et al. 2019
JUVI	PDSI pos. correlations with PPT		neg. correlation to cJun-cJul T_{mean}		Aus de Ar et a 2018
PIPO	pJul-cJune pos. correlations with PPT cApr-cAug		neg. correlation to T_{mean} in pJul, pSep, cMay, cJul		Aus de Ar et a 2018
Žofín For	est Dynamics Plot, C	zech Republic			
ABAL	no sig. correlations with June-July PPT	slight concave-down response to p.Jun-p.July <i>PPT</i> frequency	sig. pos. correlation to April T (strongest T correlation)	pos. response to Jan-March T_{max}	Kašpar, Tumajer, Vašíčková, and Šamonil, in review
FASY	no sig. correlations with June-July PPT	pos. response to p.Jun-p.July <i>PPT</i> frequency	sig. pos. correlation to Jan T (strongest T correlation)	pos. response to Jan-March T_{max}	Kašpar, Tumajer, Vašíčková, and Šamonil, in review
PIAB	modest pos. correlations (n.s) with June-July PPT >700m elev. sites moisture limited June-Aug	pos. response to p.Jun-p.July <i>PPT</i> frequency	sig. pos. correlation to March T (strongest current-year T correlation) >700m elev. sites temperature limited except June-Aug	pos. response to Jan-March T_{max}	Kašpar, Tumajer, Vašíčková, and Šamonil, in review Tumajer et al. 2017
Scotty Cr	reek, NW Territories,	Canada			Sniderhan and Baltzer 2016

Results from this study are the climate-only model. Where previous studies examined numerous climate variables or time windows (e.g., Helcoski et al., 2019), we focus on those most relevant to our findings. Beyond the methodological differences, original studies vary from this one and from one another in factors including exact set of cores analyzed, climate data sources, time frame of analysis, approaches to identifying candidate climate variables and windows (including whether this is done on a site or species level), methods for

detrending and standardizing to build chronologies, and whether the effects of temperature and precipitation are considered separately (original studies) or additively (this study). An analysis standardizing all of these factors for four species is present in Appendix ${\bf S2}$ and Fig. ${\bf S1}$.

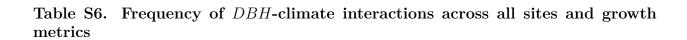


Figure S1. Comparison of our approach with traditional methods of identifying climate signals ${\bf S}$

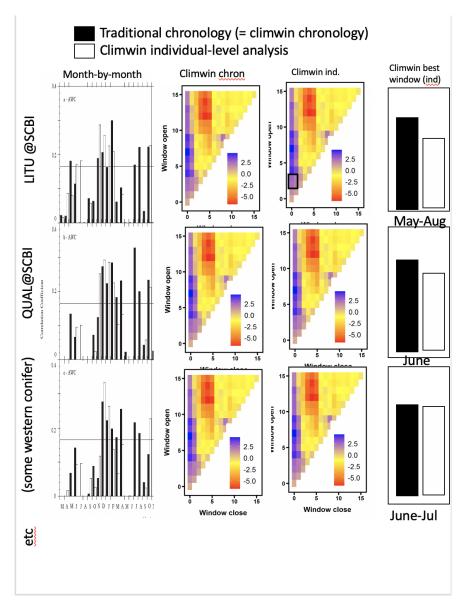


Figure S1 | (Comparison of traditional approaches with ours). (THIS FIGURE IS JUST A MOCK-UP –NOT REAL DATA. REAL FIGURE WILL INCLUDE 3-4 COMMONLY STUDIED SPECIES FROM DIFFERENT SITES.)

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