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# Herbarium specimens can reveal impacts of climate change on plant phenology; a systematic review of methods and applications

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Studies in plant phenology have provided some of the best evidence for large-scale responses to recent climate change. Over the last decade, more than thirty studies have used herbarium specimens to analyze changes in flowering phenology over time. In this review, we summarize the approaches and applications used to date. Reproductive plant phenology has primarily been analyzed using two summary statistics, the mean flowering day of year and first flowering day of year, but mean flowering day has proven to be a more robust statistic. Three types of regression models have been applied to test for changes in phenology; flowering day regressed on year, flowering day regressed on temperature, and temperature regressed on year. Most studies analyzed the effect of temperature by averaging temperatures from three months prior to the date of flowering, but other approaches may be suitable in some cases. On average, published studies have used 55 herbarium specimens per species to characterize changes in phenology over time, but in many cases fewer specimens were used. Geospatial grid data is increasingly being used for determining average temperatures at herbarium specimen collection locations, allowing testing for finer scale correspondence between phenology and climate. Multiple studies have shown that inferences from herbarium specimen data are comparable to findings from systematically collected field observations. Herbarium specimens are expected to become an increasingly important resource for analyzing plant responses to climate change. As temperatures continue to rise globally, there is a need to understand phenological rates of change in response to warming and implications of these changes, especially in tropical environments where phenological studies are thus far generally lacking.

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2 methods and applications

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47 Abstract

48 Studies in plant phenology have provided some of the best evidence for large-scale responses to recent  
49 climate change. Over the last decade, more than thirty studies have used herbarium specimens to  
50 analyze changes in flowering phenology over time. In this review, we summarize the approaches and  
51 applications used to date. Reproductive plant phenology has primarily been analyzed using two  
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56 averaging temperatures from three months prior to the date of flowering, but other approaches may be  
57 suitable in some cases. On average, published studies have used 55 herbarium specimens per species to  
58 characterize changes in phenology over time, but in many cases fewer specimens were used. Geospatial  
59 grid data is increasingly being used for determining average temperatures at herbarium specimen  
60 collection locations, allowing testing for finer scale correspondence between phenology and climate.  
61 Multiple studies have shown that inferences from herbarium specimen data are comparable to findings  
62 from systematically collected field observations. Herbarium specimens are expected to become an  
63 increasingly important resource for analyzing plant responses to climate change. As temperatures  
64 continue to rise globally, there is a need to understand phenological rates of change in response to  
65 warming and implications of these changes, especially in tropical environments where phenological  
66 studies are thus far generally lacking.

67

#### 68 Introduction

69 Carl Linnaeus pioneered the study of phenology when he outlined methods for investigating the  
70 association between flowering and climate (Puppi 2007). The word “phenology” originates from  
71 botanist Charles Morren who introduced the term around 1850 to describe his observational studies of  
72 yearly flowering (Demarée 2009). Early field studies of plant phenology have been thoroughly reviewed  
73 by van Schaik et al. (1993), Fenner (1998) and Forrest et al. (2010). Long-term field observations have  
74 provided a valuable resource for analyzing phenological responses to climate change (Walther et al.  
75 2002; Parmesan & Yohe 2003). A growing need for historical data that allows for the exploration of  
76 ecological implications of climate change has prompted researchers to look to herbarium specimens as a  
77 resource for long-term flowering data. The first study to use herbarium specimens to understand  
78 phenological responses to climate change was published by Primack et al. (2004). A few phenology  
79 studies such as Borchert et al. (1996) and Rivera & Borchert (2001) used herbarium specimens to study  
80 flowering periodicity, but not in the context of climate change. Primack et al. (2004) used 372 specimen  
81 records (1885-2002) and found peak flowering had advanced approximately eight days over the last  
82 century. Primack et al. (2004) noted that the method of using herbarium specimens may be useful for  
83 plants with either short flowering durations or long flowering durations and for plants from unique  
84 ecosystems such as mountain peaks or islands. Between 2004 and 2017, more than 30 studies were  
85 published using herbarium specimens to examine changes in phenology in response to climate change.  
86 An early criticism of the method was that plants preserved as herbarium specimens might not have been  
87 collected during their peak flowering season, potentially biasing interpretations (Lamoureux 1972). In  
88 response, authors later found that large sample sizes afforded by herbarium specimens, and the use of  
89 mean flowering times, can yield valid inferences, even if specimens were not collected at the time of  
90 peak flowering (Primack 2004; Bertin 2015). Additional criticisms of collector bias and plant size choice  
91 were also found to be overcome by appropriate statistical analyses, especially when mean flowering  
92 times were used as the variable of interest, rather than date of first flowering (Robbirt et al. 2011; Davis  
93 et al. 2015).

94 The most common approach found in studies using herbarium specimens follows the model set by  
95 Primack et al. (2004). This approach can be summarized as collecting flowering dates from herbarium

96 specimens, collecting long-term temperature data from an independent source, and then using  
97 regression analyses to analyze correlations and rates of change over time (Primack et al. 2004; Miller-  
98 Rushing et al. 2006; Gallagher et al. 2009; Robbirt et al. 2011; Gaira et al. 2011; Molnár et al. 2012;  
99 Panchen et al. 2012; Park 2012; Primack and Miller-rushing 2012; Li et al. 2013; Calinger et al. 2013; Hart  
100 et al. 2014; Rawal et al. 2015; Park & Schwartz 2015). Linear regression models are the most widely  
101 used statistical models in field studies investigating flowering phenology (Zhao et al. 2013). The studies  
102 in this review mostly used three types of linear regression models to show evidence of associations  
103 between phenology and climate change (Table 1). These studies regressed flowering day on  
104 temperature, flowering day on year, temperature on year, or some combination of these with multiple  
105 regression models. Results from these models address whether flowering day was earlier in the year  
106 when temperatures were warmer, whether flowering day was earlier in the year over time, and whether  
107 temperatures were warmer over time. Around 30% of the studies in this review used all three types of  
108 regression models in their analyses (Table 1). Approximately 82% of the studies modeled flowering day  
109 on year to show long-term changes in flowering (Table 1). About 64% of studies modeled the effect of  
110 temperature on flowering day and a different 64% modeled long-term changes in temperature (Table 1).

111 These studies have primarily been conducted with specimens from herbaria in temperate  
112 latitudes such as the Eastern Himalayas (Gaira et al. 2011; Li et al. 2013; Gaira et al. 2014; Hart et al.  
113 2014), Southern Australia (Gallagher et al., 2009; 2012; Rawal et al., 2015), Northern Europe (Robbirt et  
114 al. 2011; Diskin et al. 2012; Molnár et al. 2012), and North America (Primack et al. 2004; Lavoie and  
115 Lachance 2006; Miller-Rushing et al. 2006; Primack 2009; Neil et al. 2010; Panchen et al. 2012; Park  
116 2012; Primack and Miller-rushing 2012; Searcy 2012; Calinger et al. 2013; Park 2014; Park & Schwartz  
117 2015; Bertin 2015; Davis et al. 2015). Although studies by Borchert (1996) and Zalamea et al. (2011)  
118 analyzed flowering periodicity in tropical plants using herbarium specimens, we found no study to date  
119 that has used herbarium specimens to analyze effects of recent climate change in a tropical region. In  
120 this review, we examined how studies chose sample sizes, flowering specimens, temperature averages  
121 and geographical scale in their analyses. We also examined how these studies validated the use of  
122 herbarium specimens and provide suggestions for methods to be used in future studies.

### 123 124 Survey Methodology

125 We worked between 2015 and 2017 to compile and review studies that used herbarium specimens to  
126 assess climate change and flowering phenology. The time period of the database search was between  
127 2004 and 2015. Studies were eligible for this review if they met three main criteria. Studies must have  
128 sampled herbarium specimens, analyzed associations with climate change, and assessed flowering  
129 phenology. We searched Web of Science (1900 – present), JSTOR (1665 – present) and Google Scholar  
130 for studies containing the terms; herbarium, specimen, phenology, and climate change. We only  
131 selected studies that met the all three eligibility criteria. The methods of each study were reviewed for;  
132 sample size, determining flowering status of specimens, approach to determining temperatures,  
133 geographic variation, and any validations of the use of herbarium specimens (e.g. comparisons to field  
134 observations). Findings were then categorized into subheadings and a synthesis of each category is  
135 discussed. The reviewed papers were also categorized into studies that regressed flowering day on  
136 temperature, flowering day on year, or temperature on year and percentages were reported for each.  
137 Sample size data was also collected from the reported sample sizes of each individual study and  
138 summarized.

### 139 140 Specimen sample sizes

141 Sample size, or the number of specimens used per species, varied across studies (Table 1). The  
142 minimum number of specimens used per species was occasionally as low as two or three records (Searcy  
143 2012). Miller-Rushing & Primack (2008) used field data and found that small sample sizes led to biased

144 estimators of first-flowering dates, but mean flowering day was not biased by sample sizes. Moussus et  
145 al. (2010) investigated sample sizes by simulating 10 known phenological estimators such as mean  
146 flowering day and first-flowering date. After comparing known phenological shifts from simulated  
147 sample data with shift estimations from models using the same data, Moussus et al. (2010) concluded  
148 that first-flowering dates were inaccurate and showed much a greater variability than mean flowering  
149 day. Low sample sizes prompted Bertin (2015) to provide a detailed analysis of how sample size  
150 affected mean, median, range, early flowering and late flowering summary statistics. In random  
151 simulations comparing sample sizes, mean flowering day values deviated less than five days for species  
152 with as few as four samples (Bertin 2015). Bertin (2015) concluded that the mean was a more robust  
153 measure of phenology than other estimators of early flowering. Bertin (2015) also showed that by  
154 increasing the sample size to 20, mean flowering times deviated only one to two days.  
155 Some studies using herbarium data have set a minimum number of herbarium specimen samples per  
156 species or a minimum time range for collections to more accurately estimate phenologies and change  
157 over time. Calinger et al. (2013) and Gallagher et al. (2009) set a minimum of 10 specimens in order to  
158 meet statistical assumptions of different models. Molnár et al. (2012), eliminated a species from  
159 analyses because collections only yielded dates across an eight year time span and Park & Schwartz  
160 (2015) eliminated species with records that spanned less than three years. Neil et al. (2009) organized  
161 species into functional groups (spring ephemerals, spring shrubs, fall ephemerals, winter-spring  
162 ephemerals, and winter-spring shrubs) in order to overcome the problem of low sample size for each  
163 species. Responses of individual species within these functional groups varied greatly however and the  
164 aggregated data showed some significant trends.  
165 If phenology varies across a species' geographic range, it may be necessary have a larger number of  
166 specimens In order to analyze species distributions using herbarium specimens, van Proosdij et al.  
167 (2016) found that the minimum number of herbarium specimen samples should be between 14 and 25  
168 depending on the geographical range of the species. The van Proosdij et al. (2016) study used simulated  
169 species to assess the minimum herbarium samples required for acceptable model performance in both  
170 virtual and real study areas. Some species with narrow geographical ranges could be modeled with as  
171 few as 14 herbarium records while wide ranging species could be satisfactorily modeled with a minimum  
172 of 25 records (van Proosdij et al. 2016). Based on these studies, we recommend caution when  
173 interpreting results from samples sizes with fewer than 30 records (Miller-Rushing et al. 2008; Moussus  
174 et al. 2010; Bertin 2015). The average sample size across studies in this review was about 55 samples  
175 per species (Table 1.). We also recommend using a minimum of 10 herbarium records per species when  
176 conducting regression analyses and the mean day of year should be used rather than first flowering,  
177 especially with small samples sizes (Calinger et al. 2013; Gallagher et al. 2009).

178

179 Determining flowering status of specimens

180 Studies have often simply recorded the presence or absence of flowers on herbarium specimens.  
181 However, some studies used more detailed criteria to assess flowering specimens. Diskin et al. (2011)  
182 used a scoring system to categorize the stages of flowering on each specimen. Calinger et al. (2013) only  
183 used specimens with more than 50% of flower buds in anthesis to help ensure that the samples were in  
184 peak flowering. For a species with an inflorescence, Davis et al. (2015) only counted specimens as  
185 "flowering" if greater than 75% of flowers were open.

186 Haggerty et al. (2012) provided a primer to help phenology researchers collect data from herbarium  
187 specimens. Haggerty et al. (2012) stated that researchers must assume the stem on the herbarium  
188 sheet represents the flowering phenophase for the entire plant.

189 Studies in temperate regions have used varying methods to account for long flowering durations. For  
190 example, Molnár et al. (2012) and Bertin (2015) excluded species that flowered outside of the peak  
191 flowering season of the region, defined as the period from late-spring to early-summer. Molnár et al.

192 (2012) removed a species because its peak flowering date was in September and focused on 40 other  
193 taxa that had flowering peaks from in spring and early-summer. The excluded species was a strong  
194 outlier and it was suggested that autumn climate events may affect species differently than spring  
195 climate events (Molnár et al., 2012). Park (2012) also removed outlier records when flowering records  
196 fell outside the peak regional flowering season. Flowering records before day 45 and after day 310 on  
197 the 365 day year were removed from analyses to reduce biases caused by winter flowering species.  
198 Additionally Park (2012) removed records that were 150 days after the median flowering date for each  
199 species to reduce errors caused by any second flowerings that can happen in autumn months.  
200 Several other studies removed taxa with long flowering durations to reduce variance among species.  
201 Bertin (2015) excluded native weedy species with flowering durations from spring to fall. Gallagher et  
202 al. (2009) only used species with a flowering duration of less than three months. Panchen et al. (2012)  
203 chose to use only species with clear beginning and ending points to investigate long and short flowering  
204 duration. Panchen et al. (2012) found that plants with shorter flowering durations required smaller  
205 sample sizes to produce significant results when regressing flowering day on year. Other studies such as  
206 Calinger et al. (2013) and Lavoie & Lachance (2006) disregarded the effect of flowering duration and  
207 noted the results of Primack et al. (2004), which reported no bias associated with long or short flowering  
208 durations when mean estimations are analyzed. Plants in tropical regions often have long flowering  
209 durations (van Schaik 1993; Fenner 1998), but as long as flowering is not continuous throughout the  
210 year, methods applied to temperate regions should also yield valuable insight into effects of climate  
211 change on phenology in the tropics.

212

#### 213 Averaging temperatures

214 The foundational study by Primack et al. (2004) used temperature averages from three calendar months  
215 prior to the specimen flowering date, with the assumption that flowering date is a function of  
216 temperatures experienced in past months. Field investigations such as Fitter et al. (1995) have shown  
217 temperature averages from different sets of months preceding flowering affect flowering phenology in  
218 different ways. More recently, Calinger et al. (2013) chose to regress the month of flowering with  
219 temperature averages from each of the 11 months prior to flowering. They found that temperature  
220 averages from three months prior to the date of flowering showed the strongest correlations with  
221 flowering (Calinger et al. 2013). Robbirt et al. (2011) investigated temperature averages for three sets of  
222 temperature averages over three -month long intervals and also found that three months prior to  
223 flowering had the most predictive power. Similarly, Rawal et al. (2015) regressed flowering for each  
224 species on temperature averages from 1, 3, 6, 9, and 12 months prior to flowering, because responses  
225 can vary by species. Rawal et al. (2015) also found that mean temperatures three months prior had the  
226 greatest influence on flowering time for all species.

227 Other studies have used average temperatures from spring months because spring temperatures  
228 generally have the most predictive power for flowering date (Miller-Rushing & Primack 2008; Primack et  
229 al. 2009; Robbirt et al. 2011; Calinger et al. 2013; Park 2014; Park & Schwartz 2015). Bertin (2015) found  
230 an interesting trend that supported the effect of spring temperatures. Bertin (2015) found that the  
231 earlier a species' mean flowering time occurred in the spring, the more the species' mean dates had  
232 shifted toward an earlier day of year over time. Robbirt et al. (2011) also found the highest correlations  
233 of flowering day with spring temperature averages across March, April and May. Calinger et al. (2013)  
234 found significant changes in flowering in response to average spring temperatures (February-May) but  
235 not in response to summer temperatures (June to September). Gaira et al. (2011) found the highest  
236 correlations between flowering and temperatures in earlier months from December-February in a  
237 Himalayan perennial herb. As an alternative to using mean monthly temperatures, Diskin et al. (2011)  
238 investigated the averages of temperature anomalies, or deviations from the overall long-term mean, for

239 2, 3, and 6 month periods from January to June and found the averages from six months prior to  
240 flowering had the strongest correlations.

241 Park (2014) used temperature averages across three month periods from early spring to late summer  
242 and found a similar trend. Temperature averages were organized into early, mid, and late seasonal  
243 classes within the months of February to October. Park (2014) found warming temperatures had  
244 affected species in the early spring class more than other classes. Park & Schwartz (2015) also used  
245 early, mid and late seasonal classes for spring and summer and found that mid-season phenology events  
246 should be modeled differently than early or late season events.

247 Hart et al. (2014) used annual temperatures and temperatures from each season (spring, summer, fall,  
248 and winter) and found significant correlations for annual and fall temperature averages, but with  
249 opposite effects. Hart et al. (2014) discussed that warmer fall temperatures may delay the chilling  
250 requirement for *Rhododendron* species, resulting in a delay in flowering while warmer annual  
251 temperatures will lead to advances in flowering overall.

252 Other studies found annual temperature means were as useful as spring temperatures. Davis et al.  
253 (2015) found similar results between spring and annual temperature averages and used annual averages  
254 in analyses. Gallagher et al. (2009) also used annual temperature means for analyses and explained that  
255 seasonal means were correlated with annual means.

256 We recommend investigating the effect of temperature by analyzing averages from multiple sets of  
257 months prior to flowering for each species rather than using only spring or only annual temperatures  
258 (Diskin et al. 2011; Robbirt et al. 2011; Calinger et al. 2013). Caution should be taken when analyzing  
259 temperature averages from the same months prior to flowering for all species when flowering month  
260 varies by species. For example, when analyzing the effect of temperature averages from three months  
261 prior for all species, Calinger et al. (2013) found that for many species, flowering was correlated with  
262 temperatures three months earlier, yet for species with an earlier mean flowering day in April, January  
263 temperatures (three months prior to flowering) did not predict flowering date; instead, temperature  
264 averages from the months of February, March and April were better predictors for those species.

265

266 **Geographic variation**

267 Among species that have broad geographic ranges, differences in climate in different parts of the  
268 species' range can complicate attempts to correlate a species' flowering day with temperature. Several  
269 methods have been used to account for climate variability across a species' range. An early study by  
270 Lavoie & Lachance (2006) investigated the effects of climate variation on the phenology of Coltsfoot  
271 (*Tussilago farfara* L.) across a range of about 10,000 km<sup>2</sup> in Quebec, Canada. Temperature data from 88  
272 meteorological stations were averaged together across this range. To account for early snow cover melt  
273 in the southern part of this range, flowering dates from individuals in southern locations were  
274 normalized with individuals in northern locations by subtracting extra periods of snow cover from  
275 individuals in the north. The adjusted dates indicated flowering occurred 33 days earlier over the last  
276 century while original (unadjusted) dates indicated flowering occurred 19 days earlier over the last  
277 century.

278 While the study by Lavoie & Lachance (2006) adjusted actual dates for analyses, more recent studies  
279 mostly account for climate variation using georeferenced climate data at various scales. Calinger et al.  
280 (2013) accounted for climate variation across Ohio by using temperature averages from 10 US Climate  
281 Divisions across the state, each about 8,000 km<sup>2</sup>. A total of 344 Climatic Divisions were established  
282 across the contiguous United States in 1895 in order to monitor climate records more accurately. These  
283 divisions have now accumulated about 100 years of climate records (Guttman & Quayle 1996). A later  
284 study by Park (2014) used average temperatures across the U.S. county where each specimen was  
285 collected.



286 Other studies accounted for climate variation across longitude, latitude, or elevation. Robbirt et al.  
287 (2011) analyzed the geographical effect of longitude and found that flowering occurred 4.86 days earlier  
288 per degree of longitude in a westward direction across the southern coastal counties of England (Robbirt  
289 et al. 2011). A later study by Bertin (2015) used Hopkins' bioclimatic law to normalize dates on  
290 specimens. Hopkins' (1918) generally stated that for every increase in a degree of latitude, or increase  
291 of 121.92 m elevation, the life history events of plants and animals were delayed by four days. Bertin  
292 (2015) found consistencies with Hopkins' bioclimatic law using latitude and elevation and chose to  
293 normalize flowering dates by adding expected phenological deviations from both latitude and elevation.  
294 Gaira et al. (2011) also analyzed climate variation using elevation when temperature data were not  
295 available, assuming a 6.5°C change in temperature per 1000 m change in elevation in the Himalayan  
296 region.

297 Other studies used temperature averages across large regions. Li et al. (2013) used temperature data  
298 that was averaged from 36 meteorological stations across the Tibet Autonomous Region. Molnár et al.  
299 (2012) used temperature averages from 10 meteorological stations across Hungary and stated that the  
300 data were statistically indistinguishable across stations (~93,030 km<sup>2</sup>). Park & Schwartz (2015) averaged  
301 temperatures from 13 stations across South Carolina, USA (~82, 931 km<sup>2</sup>). A later study by Robbirt et al.  
302 (2014) used temperature averages from an area between Bristol, Preston, and London, across the  
303 United Kingdom (~17, 000 km<sup>2</sup>). Robbirt et al. (2014) used geographical divisions called Watsonian vice-  
304 counties specifically delineated for the purposes of collecting scientific data, much like the US Climate  
305 Divisions. Robbirt et al. (2014) found temperature averages were sufficient because climate variation  
306 across the Watsonian vice-counties used in their study did not significantly differ.

307 In order to more accurately estimate temperature averages across a region, recent studies have used  
308 Geographical Information Systems (GIS) to project finer-scale climate layers across a region and extract  
309 temperature data from specific Global Positioning System (GPS) points. A study by Gallagher et al.  
310 (2009) referenced GPS locations for each specimen and extracted the temperature averages at  
311 specimen GPS points from a gridded map of temperature averages across Australia (~5 km<sup>2</sup> resolution).  
312 A study by Rawal et al. (2015) also used the nearest data point from gridded climate averages across  
313 Victoria, Australia. A recent study by Edward & Still (2008) analyzed the climate envelopes of grasses by  
314 assigning GPS points to herbarium specimen locations in order to extract temperature averages from  
315 gridded climate maps (250m<sup>2</sup> resolution). Studies using GPS data are able to account for climate  
316 variation with higher resolution, although accuracy still depends on the underlying empirical data and  
317 modeling approach used to generate GIS climate layers.

318 We recommend using the most spatially precise temperature data available, such as climate divisions  
319 (Calinger et al. 2013; Robbirt et al. 2015) rather than state or region averages (Li et al. 2013; Park &  
320 Schwartz 2015). Using GPS specimen data to identify local climate conditions from GIS climate layers  
321 (Gallagher et al. 2009; Edward & Still 2008) is also now generally more precise and convenient in  
322 comparison to making generic and coarse-scale corrections for latitude, longitude or elevation (Gaira et  
323 al. 2011; Robbirt et al. 2011; Bertin 2015). If temperature averages from larger areas are used, we  
324 recommend testing for climate variability across smaller divisions before using averages across the  
325 larger area (Lavoie & Lachance 2006; Molnár et al. 2012; Robbirt et al. 2015).

326

327 Validation: herbarium specimens versus field observations

328 Field data are often combined with herbarium specimen data in analyses, allowing for comparison and  
329 sometimes allowing for validation of conclusions based on herbarium data (Primack et al., 2004; Miller-  
330 Rushing et al. 2006; Bertin 2015). Primack et al. (2004) used herbarium specimens to find a "historic"  
331 flowering mean and then used field data to find a "current" flowering mean (Primack et al., 2004). The  
332 historic mean derived from specimens was compared to the current mean from field data in order to  
333 determine if there was a statistical difference in mean flowering between the two time periods. Primack

334 et al. (2004) found that the flowering duration of each species was statistically indistinguishable  
335 between herbarium and field data and therefore concluded that herbarium and field data were  
336 compatible. Studies by Miller-Rushing et al. (2006) and Bertin (2015) also compared historical averages  
337 with current averages using herbarium specimens for historical data and field samples for current data.  
338 Miller-Rushing et al. (2006) found that results from herbarium specimens alone differed from the  
339 combined data by only about one day.

340 An early study by Borchert, (1996) found that herbarium specimen data produced slightly longer  
341 flowering durations than field data, but noted that durations were mostly similar overall. Borchert,  
342 (1996) and Rivera & Borchert (2001) found that phenology data from field sites largely overlapped that  
343 of herbarium specimens with only slight differences. The negligible differences between herbarium  
344 specimen data and field data in these studies helped justify the use of herbarium specimen data to  
345 analyze phenology in more recent studies. Nevertheless, several more recent studies specifically  
346 compared phenology estimates from field data to those made from herbarium specimens.

347 Bolmgren & Lonnberg (2005) compared herbarium specimen data directly to field data and found the  
348 two data sets were overall highly correlated with only minor differences. For example, herbarium  
349 specimens showed a slightly earlier mean flowering for spring-flowering plants than field data, but the  
350 difference was not significant (Bolmgren & Lonnberg, 2005). Later studies by Robbirt et al. (2011) and  
351 Davis et al. (2015) also primarily focused on testing the validity of using herbarium specimen data.  
352 Robbirt et al. (2011) used a principal axis regression analysis to compare herbarium derived peak-  
353 flowering dates with field derived peak-flowering dates and found a high degree of correlation. Robbirt  
354 et al. (2011) discusses how the high degree of correlation between herbarium and field data also  
355 supports the notion that geographically different records will not significantly alter the robustness of  
356 either data set. A study by Davis et al. (2015) used a paired t test to compare mean first flowering day  
357 between herbarium specimen and field data and found no statistical difference. Davis et al. (2015)  
358 concluded that both specimen and field data could be combined and used as a whole.

359 In order to increase sample sizes, Molnár et al. (2012) added about 2000 field observations to about  
360 5000 herbarium records, resulting in 70% herbarium records for the study. Similarly, Panchen (2012)  
361 added about 2000 field records to about 1500 herbarium records, for a total of 43% herbarium records  
362 for the study. Searcy (2012) combined herbarium specimen and field data and then split the combined  
363 data into two time periods (1863–1935 and 1994–2008). Despite criticisms, herbarium specimen data  
364 have been shown to produce similar enough results to field data that herbarium specimen data are now  
365 widely accepted in phenological studies.

366

## 367 Conclusions

368 The use of herbarium specimens for the investigation of flowering phenology has grown considerably  
369 during the past decade. As efforts to produce digital copies of specimens and label information have  
370 amassed large datasets, new approaches for analyzing responses to climate change are becoming  
371 available. Studies using herbarium specimens have become an asset for long-term climate change  
372 vulnerability assessment. Studies using herbarium specimens have also begun to analyze the effects of  
373 climate change on community composition (Miller-Rushing & Primack 2008, Park 2014) coevolved plant  
374 pollinator relationships (Molnár et al. 2012; Robbirt et al. 2014) functional groups (Miller-Rushing &  
375 Primack 2008; Panchen et al 2012; Calinger et al. 2013) and phylogenetic relationships (Bolmgren &  
376 Lonnberg 2005; Molnár et al. 2012; Primack and Miller-rushing 2012).

377 While studies using herbarium specimen data to analyze long-term changes have been limited to  
378 temperate regions, future studies could use circular statistics to analyze long-term phenological changes  
379 in tropical regions (Fisher 1993; Morellato et al. 2010). Circular statistics have been used to analyze  
380 flowering phenology in several tropical field studies, but these studies lacked long-term climate change  
381 analyses (Novotny & Basset 1998; Morellato et al. 2000; Cruz et al. 2006; Rogerio & Araujo 2010;

382 Tesfaye et al. 2011; Nadia et al. 2012; Nazareno & dos Reis 2012; Staggemeier et al. 2012; Carvalho &  
383 SartorWe2015; Kebede & Isotalo 2016). Although small sample sizes have been used in early studies of  
384 phenology, various factors, such as long flowering duration or wide geographic range, may require larger  
385 sample sizes. Based on recent validations, mean estimations of peak flowering should be used rather  
386 than first flowering because estimates of first flowering from herbarium specimens are more variable.  
387 The use of GPS data appears to be the way forward for the advancement of methods in the study of  
388 phenology. GPS point data will allow for correspondence with higher resolution temperature data in  
389 climatically diverse geographical regions. Studies using herbarium specimen data will continue to help  
390 us understand the impact of recent climate change on plant reproductive phenology. Future studies are  
391 needed on the importance of phylogenetic signals and plasticity in order to further improve our  
392 understanding of adaptation and resilience to climate change, and herbarium specimens offer promise  
393 in understanding effects of climate on other aspects of plant phenology such as fruit ripening and spring  
394 leaf emergence, the timing of which has important implications for higher trophic levels, which may  
395 include rare animals dependent on plant resources.

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**Table 1** (on next page)

Table1

Methods of studies; The column "Flw Day ~ Temp" represents studies that conducted a type of regression analysis with flowering day (Flw Day) as the dependent variable and temperature average (Temp) as the independent variable. This follows for columns using a tilde (~) which include the independent variable "Year". The " $\Delta \bar{x}$ " symbol represents studies that analyzed a difference in the mean flowering day between Historic and Current time period groups rather than using a type of regression analysis.

species	specimen		authors	year	geographic			
	specimens	per species			region	(flw day ~ temp)	(flw day ~ year)	(temp ~ year)
1	117	117	Gaira et al.	2011	Eastern Himalayas		x	x
1	NA	NA	Gaira et al.	2014	Eastern Himalayas	x	x	x
1	192	192	Robbirt et al.	2011	Northern Europe	x		
5	158	32	Rawal et al.	2015	Southern Australia	x	x	
5	540	108	Diskin et al.	2012	Northern Europe	x	x	x
20	371	19	Gallagher et al.	2009	Southern Australia	x	x	x
20	1108	55	Davis et al.	2015	North America	x	x	x
28	1587	57	Panchen et al.	2012	North America	x	x	
36	460	13	Hart et al.	2014	Eastern Himalayas	x		x
>37	372	10	Primack et al.	2004	North America	x	x	x
39	216	6	Lavoie & lachange	2006	North America		x	x
39	5424	139	Molnár et al.	2012	Northern Europe		x	x
41	909	22	Li et al.	2013	Eastern Himalayas	x	x	x
42	142	3	Miller-Rushing et al.	2006	North America	x	x	x
43	NA	NA	Primack & Miller-Rushing	2012	North America		x	x
87	NA	NA	Neil et al.	2010	North America		x	
141	5053	36	Calinger et al.	2013	North America	x		x
186	30,000	161	Bertin	2015	North America		$\Delta \bar{x}$	x
370	1125	3	Searcy	2012	North America		$\Delta \bar{x}$	
1185	5949	5	Park	2012	North America		x	
>1700	19,328	11	Park	2014	North America	x		
24,105	823,033	34	Park & Schwartz	2015	North America	x	x	