

## High temperatures reduce growth, infection, and transmission of a naturally occurring fungal plant pathogen

By: Dalia V. Chen<sup>1</sup>, Samuel P. Slowinski<sup>1</sup>, Allyson K. Kido<sup>1,2</sup>, Emily L. Bruns<sup>1</sup>

<sup>1</sup>Biology, University of Maryland at College Park, College Park, Maryland, USA

Corresponding author: Emily L. Bruns, [ebruns@umd.edu](mailto:ebruns@umd.edu)

Journal: Ecology- Article

Open research statement: All of the supporting data and R code is archived at Dryad:

10 doi:10.5061/dryad.4mw6m90jv

11

12 Key words: Climate change; disease ecology; heat curing, infection; *Microbotryum*; *Silene*

### 13 *latifolia*, thermal tolerance; transmission

14

15

16

17

18

10

20

21

2

24 **Abstract**

25 Climate change is rapidly altering the distribution of suitable habitats for many species as well as  
26 their pathogenic microbes. For many pathogens, including vector-borne diseases of humans and  
27 agricultural pathogens, climate change is expected to increase transmission and lead to pathogen  
28 range expansions. However, if pathogens have a lower heat tolerance than their host, increased  
29 warming could generate ‘thermal refugia’ for hosts. Predicting the outcomes of warming on  
30 disease transmission requires detailed knowledge of the thermal tolerances of both the host and  
31 the pathogen. Such thermal tolerance studies are generally lacking for fungal pathogens of wild  
32 plant populations, despite the fact that plants form the base of all terrestrial communities. Here,  
33 we quantified three aspects of the thermal tolerance (growth, infection, and propagule  
34 production) of the naturally occurring fungal pathogen *Microbotryum lychnidis-dioicae*, which  
35 causes a sterilizing anther-smut disease on the herbaceous plant *Silene latifolia*. We also  
36 quantified two aspects of host thermal tolerance: seedling survival and flowering rate. We found  
37 that temperatures  $>30$  °C reduced the ability of anther-smut spores to germinate, grow, and  
38 conjugate *in vitro*. In addition, we found that high temperatures (30 °C) during, or shortly after  
39 the time of inoculation strongly reduced the likelihood of infection in seedlings. Finally, we  
40 found that high summer temperatures in the field temporarily cured infected plants, likely  
41 reducing transmission. Notably, high temperatures did not reduce survival or flowering of the  
42 host plants. Taken together, our results show that the fungus is considerably more sensitive to  
43 high temperatures than its host plant. A warming climate could therefore result in reduced  
44 disease spread or even local pathogen extirpation, leading to thermal refugia for the host.

45

46 **Introduction**

47 Earth's climate is warming faster than many climate models initially predicted (IPCC 2022), and  
48 these changes in temperature are likely to shift species interactions (Gilman et al. 2010),  
49 including host-parasite interactions (Budria and Candolin 2014). This is because temperature  
50 plays a critical role in the development, expression, and timing of both host and pathogen life  
51 history traits (Verant et al. 2012, Mordecai et al. 2013, Shocket et al. 2018). Indeed there is now  
52 a growing body of work showing that warming temperatures are likely to increase transmission  
53 and the geographic distribution of many important vector-borne human pathogens (Mordecai et  
54 al. 2017, Shocket et al. 2020) and pathogens of agricultural crops (Rossi et al. 2001, Bebber et al.  
55 2013). However, not all pathogens will necessarily benefit from a warmer climate. For example,  
56 the invasive fungal pathogen of amphibians, *Batrachochytrium dendrobatidis* (*Bd*), is  
57 particularly sensitive to high temperatures (Nowakowski et al. 2016). A useful, recent framework  
58 for predicting organismal responses to climate change is the thermal mismatch hypothesis  
59 (Cohen et al. 2017). If pathogens have a lower thermal maximum (the maximum temperature for  
60 survival and growth and reproduction) than their hosts, then warming temperatures can increase  
61 the disease-free niche space for the host, generating 'thermal refugia' for hosts (Gsell et al.  
62 2023). However, predicting thermal refugia requires knowledge of the thermal optima and  
63 maximum of both host and pathogen (Mordecai et al. 2013).

64 Plants form the base of all terrestrial food chains; understanding and predicting how  
65 climate change will affect plant-pathogen interactions is critically important. Yet, to date, most  
66 measures of plant-pathogen thermal tolerance have focused on agricultural pathogens (Rossi et  
67 al. 2001, Raftoyannis and Dick 2002, Chaloner et al. 2021, Shah and Bergstrom 2000), where  
68 there is growing evidence that warming will exacerbate disease spread (Bebber et al. 2013). In

69 contrast, much less is known about the thermal tolerance of wild plant pathogens (Robin et al.  
70 2017), despite the fact that pathogens can have strong impacts on plant population growth  
71 (Alexander 2010, Gilbert 2002), alter competitive interactions (Mordecai 2011), and affect  
72 community composition (Bever et al. 2015). In many cases, only *in vitro* pathogen growth rates  
73 are measured (Robin et al. 2017), which often have a wider thermal tolerance range than *in*  
74 *planta* traits (Chaloner et al. 2021), or knowledge of pathogen thermal tolerance is inferred from  
75 field distribution data (Penczykowski et al. 2015, Scala et al. 2019, Dudney et al. 2021).

76 *Microbotryum lychnidis-dioicae* is a naturally occurring fungal pathogen of the  
77 herbaceous plant *Silene latifolia* and has become a model system for disease ecology  
78 (Bernasconi et al. 2009). Infected plants produce spores instead of pollen, resulting in host  
79 sterility. Several lines of evidence suggest that *M. lychnidis-dioicae* has a lower heat tolerance  
80 than its host. First, while the host plant *Silene latifolia* occurs across a broad range of latitudes  
81 and elevations, disease prevalence is higher in cooler climates and at higher elevations (Hood et  
82 al. 2010, Abbate and Antonovics 2014, Bruns et al. 2018, Kido and Hood 2020). Second, in  
83 greenhouse settings, high heat has been observed to cause temporary host recovery, enabling the  
84 host to produce fertile flowers (Elmqvist et al. 1993). However, a controlled study of the effect  
85 of temperature on infection, growth and transmission ability of this model plant-pathogen system  
86 has never been carried out. Furthermore, it is unclear whether the ‘heat-curing effect’ is limited  
87 to a greenhouse environment or also occurs in natural field settings.

88 Here we used a combination of lab, greenhouse, and field experiments to measure the  
89 relationship between temperature and the following three key life history stages of *M. lychnidis-*  
90 *dioicae*: *in vitro* performance of the spores, seedling infection ability, and the ability to sporulate  
91 in the host’s anthers. We also quantified host survival and reproductive ability across these same

92 temperatures. We hypothesized that high temperatures would reduce the overall ability of the  
93 pathogen to germinate, infect and reproduce within its host, without reducing host survival and  
94 reproduction. Our results show that temperatures  $> 30^{\circ}\text{C}$  strongly reduced all three measures of  
95 pathogen performance. However, host survival and flowering ability did not decline at high  
96 temperatures, suggesting that climate change could substantially alter host-pathogen interactions  
97 in ways favorable to the host.

98

## 99 **Materials and Methods**

100

### 101 **Study System**

102 *Silene latifolia* (formerly *Silene alba*; Caryophyllaceae) is an herbaceous short-lived  
103 perennial plant native to Europe. It was introduced to North America in the mid 1800s and is  
104 now found across the east coast (Taylor and Keller 2007), where it tends to grow in patchy  
105 roadside metapopulations (Alexander and Antonovics 1988). The plant is dioecious; individual  
106 plants are either pollen-producing males or seed-producing females.

107 *Microbotryum lychnidis-dioicae* (Denchev et al. 2008) is a fungus in the Basidiomycota  
108 that causes a systemic anther-smut disease on *Silene latifolia*. It is a naturally occurring endemic  
109 pathogen in both the native range in Europe and in North America where it was introduced  
110 (Fontaine et al. 2013). The fungus hijacks the plant's flowers for its own reproduction, producing  
111 diploid teliospores in the anthers, and preventing pollen production. Infected female plants also  
112 produce spore-filled anthers, and the ovary is sterilized (Schäfer et al. 2010). Infection does not  
113 cause plant mortality, but the fungus is systemic and can spread throughout the plant, potentially  
114 affecting all flowers, rendering plants completely sterile. Transmission of the spores to new hosts

115 can occur through pollinator movement and highly localized wind transmission (Bruns et al.  
116 2017, Roche et al. 1995). Before teliospores of *M. lychnidis-dioicae* can infect *S. latifolia*, they  
117 germinate and go through meiosis producing haploid sporidia that can reproduce asexually by  
118 budding. Each sporidium is one mating type, either a1 or a2. Prior to infection, two sporidia of  
119 opposite mating types must conjugate to produce an infectious hypha (Giraud et al. 2008).

120

## 121 **Overview of lab and field experiments**

122 We carried out three different experiments to determine the effect of temperature on the  
123 expression of *M. lychnidis-dioicae* and *S. latifolia* life-history traits. In Experiment 1, we assayed  
124 metrics of teliospore germination, haploid sporidial growth, and conjugation rates *in vitro* under  
125 a range of temperatures from 14-30 °C. In Experiment 2, we measured pathogen infection rate of  
126 *S. latifolia* seedlings, and survival of uninoculated seedlings under three temperatures (24, 26,  
127 and 30 °C) with a greenhouse experiment. In Experiment 3, we measured the expression of  
128 disease in infected adult plants in a field setting over a six-month period and analyzed the  
129 relationship between average high temperature and disease expression.

130

### 131 **Experiment 1: Effect of temperature on spore traits *in vitro***

132 Isolating strains and mating types: Three strains of *M. lychnidis-dioicae* were isolated from a  
133 single infected population of *Silene latifolia* in June 2020 in Giles Co, VA (37.32867 N -  
134 80.48812 W). Each strain used in this study was isolated from a single infected bud, each from a  
135 separate plant. Dilution plating was used to generate two haploid sporidial lines from each strain:  
136 one a1 mating type line and one a2 mating type line (see Appendix\_S1 for details). In total, this

137 process produced six sporidial cultures: both the a1 and a2 mating types for three original  
138 infected plants.

139

140 **Pathogen response to temperature *in vitro***

141 We measured three pathogen traits *in vitro*: teliospore germination, sporidial growth, and  
142 conjugation. All traits were assayed at 14 different temperatures (every 2 °C from 4-30 °C) in  
143 three Percival incubators (model I30BLLC8) with a 16-hour day and 8-hour night. Nighttime  
144 temperatures were set at 2 °C lower than the day temperature. Three temperatures were assayed  
145 in each experimental run, and experimental runs were repeated with different combinations of  
146 assay temperatures (order randomly determined) until every temperature had been assayed in two  
147 experimental runs. At each temperature within each experimental run there were three technical  
148 replicates (i.e., different plates) of each strain.

149 *Teliospore Germination*: We collected fresh teliospores from infected flowers within a  
150 common garden experiment in Beltsville, MD (see Experiment 3 below for details). Infected  
151 plants in the garden were originally inoculated with a mix of all three *M. lychnidis-dioicae* stains.  
152 Teliospores were diluted in sterile DI water to a concentration of 500 spores/µL, and 200 µL of  
153 this solution was then transferred onto a 1.5% water agar plates. Plates were incubated for 20  
154 hours before imaging at 10X magnification on a light microscope (Leica DMI1). The number of  
155 germinated teliospores were counted manually in five fields of view on a Leica MC170 camera.

156 *Colony growth*: Sporidial viability and growth was assayed as the number of individual  
157 sporidial colonies visible after 7 days on 1% PDA (Potato Dextrose Agar; Difco<sup>TM</sup>). Each  
158 sporidial line was diluted in sterile DI water to a concentration of 1 sporidia/ µLand 100 µLs

159 were plated onto 1% PDA (approximately 100 sporidia per plate). Plates were incubated for one  
160 week and then the number of visible colonies were counted manually.

161 Conjugation rate: For each strain, a solution was made containing a 50:50 mix of both a1  
162 and a2 haploid sporidia at concentration of 100,000 spores/ $\mu$ L suspended in double deionized  
163 sterile water. Six droplets of the conjugation mix were pipetted onto a detached *S. latifolia* leaf  
164 suspended on a 1.5% water agar plate and incubated for 20 hours. After the incubation droplets  
165 from a single leaf were combined onto a glass slide with 1  $\mu$ L of Lactophenol-Cotton Blue stain  
166 and a coverslip. Slides were imaged at 40X magnification on a light microscope (Lecia DM1000  
167 LED) with a Leica MC170 HD camera. The number of single and conjugating sporidia was  
168 counted manually across 4 images per slide.

169

## 170 **Statistical methods, Experiment 1**

171 All statistical tests were run in R version 4.1.2. We used generalized additive models  
172 (GAM, package ‘mgcv’; Wood 2011) to model the relationship between temperature and the  
173 three *in vitro* traits measured in Experiment 1, because this approach allowed more precise  
174 estimates of the non-linear relationship. Temperature replicate was included as a random factor  
175 in all models. To avoid overfitting, we varied the number of knots and visually inspected the  
176 change in fit. We then chose the minimum knot number that showed a smoothed, single peaked  
177 curve that did not substantially change with additional knots (See Appendix S1: Figure S1).

178 For growth, we included mating type and strain as additional linear predictors in the  
179 model. For conjugation rate, strain was included (mating type was not because both mating types  
180 were required for each conjugation event). Strain was not included in the germination model

181 because the teliospores were of unknown strain. We tested each effect in the model with  
182 likelihood ratio tests.

183 We used the ‘predict’ function to generate predicted trait values and 95% prediction  
184 confidence intervals for the measured temperature interval between 4-30 °C. The lower and  
185 upper temperature thresholds for each trait were defined as the temperature where the lower 95%  
186 confidence interval first went below zero on both sides of the temperature curve. The optimum  
187 temperature was defined as the temperature at which the focal trait had the highest predicted  
188 value.

189

## 190 **Experiment 2: The effect of temperature on infection rate of seedlings**

191 To test whether high temperatures resulted in reductions in infection *in planta* we set up a  
192 seedling inoculation experiment. We used incubators to manipulate the temperature at  
193 inoculation or four days after inoculation. In total we had five treatment groups: control-24 °C,  
194 26 °C immediately at inoculation, 26 °C four days delayed, 30 °C immediately at inoculation, 30  
195 °C four days delayed. For the immediate treatments the seedlings were heated to their treatment  
196 temperature for four days starting at inoculation and then moved to a 24 °C incubator. For the  
197 delayed treatments the seedlings were kept at 24 °C starting at inoculation for four days then  
198 moved to the treatment temperatures for four days. Within each of the five temperature  
199 treatments we had an inoculated and mock-inoculated group. The mock-inoculated group was  
200 used to assess the impact of temperature on the host plant in the absence of disease.

201 Seeds were planted in 148mm culture dishes (Falcon) containing 1% water agar. We used  
202 seeds from four different greenhouse reared full-sib families of *S. latifolia*, originally collected  
203 from Giles, Co. VA. These families were chosen because they previously had been shown to be

204 highly susceptible to anther-smut at the seedlings stage (Slowinski et al., unpublished data). Each  
205 culture dish contained 30 seeds of each family (120 seeds per dish) and we planted 3 dishes per  
206 temperature-treatment\*inoculation-treatment (4 families \* 30 plants per dish \* 3 replicate dishes  
207 \* 5 temperature treatments \* 2 inoculation treatments = 3,600 seeds). Culture dishes were placed  
208 at a 12-hour day/night 24 °C/ 22 °C cycle in Percival incubators (model I30BLLC8) for one  
209 week.

210 Inoculations were carried out 7 days after planting. Incubator temperatures were adjusted  
211 to treatment temperatures the day of inoculation. We diluted fresh sporidia in distilled water  
212 from each of the three *M. lychnidis-dioicae* strains (a1 and a2) used in Experiment 1 to a  
213 standard concentration of 5,000 sporidia/µL. These were combined in equal proportions to make  
214 a single inoculation mixture of both a1 and a2 mating types of all three strains. We applied 3 µL  
215 of inoculation solution to the apical meristem of each seedling in the inoculation treatment.  
216 Seedlings in the mock-inoculation group were inoculated with 3µL of sterile double-deionized  
217 water.

218 On day 15, plates were removed from the incubators and 15 plants per family per plate  
219 (total 1,800 seeds) were transplanted into 1.5 x 8.25" Cone-tainers (Stuewe and Sons Inc) filled  
220 with Sunshine Mix #1 (Sungrow Horticulture). Cones were placed in racks in a blocked  
221 randomized design. After transplanting, the seedlings were kept in a greenhouse mist room for 2  
222 weeks to allow the young plants to root, then moved into a greenhouse with 16-hour days. Plants  
223 were reared to flowering and then scored for presence of spores in their anthers. We also  
224 recorded any mortality, classified as plants with no green leaves left.

225

226 **Statistical Methods, Experiment 2**

227 To determine whether temperature affected infection rate, we used a generalized linear  
228 model with a binomial error structure. We included temperature treatment, plant family and  
229 replicate culture dish as linear predictors. Only plants in the inoculated treatment were included.  
230 Since initial model checking indicated overdispersion we used a quasibinomial model with a  
231 dispersion parameter of 1.8345. We first tested for an interaction effect between  
232 temperature\*family, using a likelihood ratio test to compare the fit of models with and without  
233 the interaction effect. The interaction was not significant and was dropped from the final model,  
234 which included only main effects (see results). We visually checked the final model to confirm  
235 assumptions of homoscedasticity and normality. We tested the significance of each predictor in  
236 the final model using likelihood ratio tests.

237 To determine which of the temperature treatments differed significantly from the control  
238 treatment of constant 24 °C, we ran a one-way ANOVA on the proportions and followed up with  
239 a post-hoc Dunnett's test.

240 To determine whether temperature treatment affected seedling survival, we used a  
241 binomial glm on the whole data set, with temperature, inoculation treatment, plant family, and  
242 replicate culture dish as linear predictors. We started with a full model that included main effects  
243 and all pairwise interactions and removed any non-significant interaction terms (see results). We  
244 tested the significance of each predictor in the model using likelihood ratio tests.

245

### 246 **Experiment 3: Effect of temperature on disease expression of infected plants in the field**

247 We inoculated *Silene latifolia* plants with a mixture of the same three *Microbotryum*  
248 strains used in experiments 1 and 2 and reared them in the greenhouse. When the plants  
249 flowered, we discarded the healthy plants and selected 178 infected plants and transplanted them

250 into a common garden at the Maryland Agricultural Experiment Station (MAES) in Beltsville,  
251 Maryland. Infected plants were transplanted into 1-gallon pots filled with Sunshine mix 160  
252 (Sungrow Horticulture). Pots were placed in 15 rows (spaced 3m apart), with 12 pots per row  
253 (spaced 3.75m apart). Each pot was sunk 6 inches into the soil and hooked up to an irrigation  
254 dripline. Pots were placed into the field on May 17, 2021. Every two weeks from May 17, 2021  
255 until November 18, 2021 each plant was scored as flowering or not, as well as for the presence of  
256 spore-producing flowers. A local weather station at MAES collected daily information on the  
257 local maximum and minimum temperatures.

258

### 259 **Statistical Methods: Experiment 3**

260 Out of the 178 infected plants placed into the field, 6 never produced infected flowers and  
261 were excluded from the analysis. We defined the proportion infected as the proportion of plants  
262 producing at least 1 infected flower out of the total number flowering during a particular time  
263 point. We calculated the average high and low temperatures for the 7, 10, or 14-day period prior  
264 to each data collection time point using the MAES weather station data. To determine which  
265 temperature time-window (7 to 14-day average) was the best predictor of disease expression, we  
266 fit individual GAMs (Wood 2011) with a single temperature time-length as a smoothed predictor  
267 and compared proportion of deviance explained and the Un-biased Risk Estimator (UBRE;  
268 Wood 2006). We found that a 14-day average high temperature explained the most deviance and  
269 had one of the lowest UBRE scores (Appendix S1: Table S2).

270 Once we identified the best-fit temperature time window, we used this to determine the  
271 ‘cutoff’ high temperature that resulted in a decrease in infected flowers (i.e., an increase in  
272 ‘curing’). To do this, we broke the data set into sets at a cutoff temperate and fit two lines (using

273 ordinary linear regression) for the data set before and after the cutoff. We recorded the total  
274 residual standard error across both models. Then we repeated this process with a range of cut-off  
275 temperatures and found the cutoff temperature that minimized the total residual standard error.

276 To determine how high heat impacted host plant flowering rate, we used linear regression  
277 to assess the relationship between the average 14-day high temperature and proportion of plants  
278 flowering.

279

## 280 **Results**

### 281 **Experiment 1: Effect of temperature on spore traits *in vitro***

282 Temperature significantly affected germination rate, growth of colonies after one week,  
283 and conjugation rate (Table 1, Figure 1). All traits declined at high temperatures ( $>28$  °C).  
284 However, traits varied in their sensitivity to heat, with the upper temperature threshold for  
285 germination rate  $>$  growth rate  $>$  conjugation rate (Table 1). Pathogen strain had a significant  
286 effect on colony growth ( $df=2$ ,  $F=6.418$ ,  $p=0.0018$ , Appendix S1: Table S1) and a marginally  
287 significant effect on conjugation rate (Appendix S1: Table S1). Estimates of thermal maxima  
288 were robust to variation in the number of knots used in the gam fits (Appendix S1: Figure S1).

289

### 290 **Experiment 2: The effect of temperature on infection rate of seedlings**

291 A total of 706 plants flowered and were scored for infection. Of these, 278 (39%) were  
292 infected. In the mock inoculation group, a total of 406 plants flowered and 2 plants became  
293 infected (0.5%).

294 Temperature treatment and family had a significant effect on the proportion of plants that  
295 became infected (Table 2). There was no significant temperature\*family effect (Deviance =

296 20.434, df = 12, p=0.466). Infection was highest in the control treatment (constant 24 °C) and  
297 showed a general pattern of decline with increasing temperatures (Figure 2). A one-way  
298 ANOVA, followed by a Dunnett's test found that only the two hottest treatments (30 °C  
299 immediate and four days delayed) were significantly different from the control (Figure 2,  
300 Appendix S1: Table S3). The 30 °C immediate treatment had a 64% reduction in infection  
301 compared to the control.

302 A total of 173 seedlings died (9.6%) within 5 months of receiving the temperature  
303 treatments. Seedling mortality was significantly affected by inoculation treatment and family, but  
304 not by temperature (Table 3; Appendix S1: Figure S2). Mortality was higher in the inoculated  
305 treatment (weighted average and 95% CI =12.3% +/- 0.016) than in the mock-inoculated  
306 treatment (6.89% +/- 0.021). There were no significant interaction effects, as evidenced by the  
307 loglikelihood comparison between a model with all pairwise interactions and a model with only  
308 the main effects shown in Table 3 (Deviance = 21.314, df =31, p=0.903).

309

### 310 **Experiment 3: Effect of temperature on disease expression of infected plants in the field**

311 Temperature varied significantly over the course of the season, with significant heat  
312 waves in July and August (Figure 3a). Plants flowered continuously over the season, with the  
313 highest number in June and July (Figure 3b). Plants began to 'cure' and lose their infected  
314 flowers in July, with symptomatic infections returning in September (Figure 3b).

315 Temperature, measured as the 14-day average, had a significant positive, linear effect on  
316 the number of plants flowering (slope =8.693, se=1.187, df = 25, t=7.326, p<0.0001, Figure 3c).

317 Temperature had a non-linear effect on disease expression. Expression was not strongly  
318 impacted by increasing temperatures up to a threshold close to 27°C, after which increasing

319 temperature led to a rapid loss of disease symptoms (Figure 3d). This cutoff value was confirmed  
320 by the two-line linear model, which was optimized at a tipping point between 26.5-27.2°C  
321 (Appendix S1: Figure S3, Table S4).

322

### 323 **Discussion**

324 Our results show that temperatures greater than 30 °C, an ecologically relevant  
325 temperature for *S. latifolia*, can strongly limit the germination, growth, and conjugation of the  
326 fungus *in vitro*, reduce infection success on seedlings, and result in temporary curing of infected  
327 adult hosts. Moreover, host survival and flowering are not impeded at 30 °C, indicating a thermal  
328 mismatch in the host's favor. This low heat tolerance of *Microbotryum* could help to explain why  
329 the disease is more prevalent on *S. latifolia* at higher elevations and latitudes where the climate is  
330 cooler (Abbate and Antonovics 2014, Bruns et al. 2018, Hood et al. 2010). Our results also  
331 suggest that increasing global temperatures could limit the future distribution of the disease.

332

#### 333 Thermal tolerance of pathogen and host

334 From our *in vitro* experiment we found that three pathogen traits critical to infection  
335 (germination, growth, and conjugation) had a non-linear relationship with temperature, similar to  
336 other pathogens in plants and animals (Mboup et al. 2012, Mordecai et al. 2017, Scala et al.  
337 2019, Ragonese et al. 2024). All three traits were strongly suppressed at temperatures  
338 approaching 30 °C. However, conjugation had the lowest heat tolerance, with an upper threshold  
339 of 25 °C, suggesting this may be the temperature-limiting trait for infection.

340 Our experiments on live plants indicated that this observed *in vitro* heat sensitivity of the  
341 pathogen can have important consequences for infection success and transmission *in vivo*. High

342 temperatures reduced the probability that exposed seedlings would become infected, in some  
343 cases by as much as 64%. In both the 26 and 30 °C heat treatments, the reduction in infection  
344 tended to be greater for seedlings that were heat-treated immediately compared to four days  
345 delayed, indicating that the time window following exposure during which high temperatures can  
346 prevent infection is limited. However, at 30 °C there was still a significant reduction in infection  
347 even in the delayed heat treatment. Schäfer et al. (2010) observed infection structures on the leaf  
348 surface four days post-inoculation and visualized the fungus in the intercellular spaces within the  
349 plant 8 days post-inoculation, but there could be variation in this timing. Thus, it is unclear  
350 whether the reduction in infection in the delayed 30 °C treatment was due to reduced survival  
351 and conjugation on the leaf surface or reduced ability to infect and survive within the plant.

352         Perhaps most critically, we found that 2-week periods of high temperature (>27 °C)  
353 temporarily cured infected adult plants in the field, enabling hosts to produce healthy flowers and  
354 reducing the total duration of sporulation and transmission. The symptom-curing effects and the  
355 new infection suppression effects of heat are likely to have compounded effect on transmission  
356 in the field, with high heat leading to both a reduction in the number of infectious spores landing  
357 on seedlings and a reduced infection rate for the spores that are produced.

358         While we found a strong suppressive effect of heat on the anther-smut pathogen, the plant  
359 host was much more tolerant of high temperatures. We found no significant effect of temperature  
360 on seedling mortality in the growth chamber, and heat had a strong positive effect on host  
361 flowering in the field. These results indicate a mismatch in the thermal tolerance of pathogen and  
362 host (Cohen 2017), with low heat tolerance in the pathogen but a more heat-tolerant host. Taken  
363 together, our findings indicate that global warming could generate ‘thermal refugia’ (Gsell et al.

364 2023), regions of the host range where the warm temperatures significantly reduce transmission  
365 rates, or even extirpate the disease locally, releasing the host population from pathogen pressure.

366 Our study adds to a growing body of work that shows a high diversity of thermal  
367 tolerances among pathogens. For many pathogens, especially human vector-borne pathogens  
368 (Mordecai et al. 2017), aquatic pathogens (Ward et al. 2007, Shocket et al. 2018), and well-  
369 studied agricultural pathogens (Rossi et al. 2001, Bebber et al. 2013, Chaloner et al. 2021),  
370 transmission is favored by warmer temperatures, leading to strong concerns that global warming  
371 will exacerbate the threats of disease (Tracy et al. 2019, Chaloner et al. 2021, Stukenbrock and  
372 Gurr 2023). In a zooplankton-fungal disease, warmer temperature have been experimentally  
373 shown to increase transmission (Shocket et al. 2018). Among pathogens of natural plant  
374 populations, manipulative warming experiments in alpine meadows have been shown to increase  
375 the overall level of foliar pathogens in the community (Liu et al. 2019). Additionally, warmer  
376 winters have been linked to increased survival of powdery mildew pathogens (Penczykowski et  
377 al. 2015).

378 In contrast our results show that not all diseases are likely to increase with warming  
379 temperatures. Similar heat sensitivity has also been reported for a few other pathogens such as a  
380 protozoan parasites of monarch butterflies, where infection probability declines at higher  
381 temperatures, however the host also suffers at higher temperatures (Ragonese et al. 2024)..  
382 Among fungal pathogens of wild plants heat sensitivity has also been suggested by long term  
383 observational studies: For example, Zhan et al. (2018) showed the local extinction rate of the rust  
384 fungus *Triphragmium ulmariae* within a Swedish metapopulation of its host, *Filipendula*  
385 *ulmaria* has increased with warming over a 26-year period. Likewise, Dudney et al. (2021) used  
386 a 20-year data set to show that white pine blister rust (caused by the fungus *Cronartium ribicola*)

387 has moved up to higher, cooler elevations and been extirpated lower elevations in the Sierra  
388 Nevada. Our study contributes to these findings by showing that multiple components of  
389 pathogen life history are inhibited by high temperatures.

390

391 *Ecological and evolutionary consequences*

392 Our results indicate that climate change could strongly affect the seasonality of anther  
393 smut transmission. Longer and more intense periods of summer heat could limit transmission to  
394 shorter peaks in the spring and fall, as infected plants cure during the heat of summer. This  
395 shorter seasonal duration of transmission could lower prevalence and potentially drive local  
396 pathogen extinction. Alternatively, shorter and warmer winters could increase the suitable  
397 periods for flowering and sporulation in the spring and fall, offsetting the loss of summer  
398 transmission. Continued phenological measurements of both the host and pathogen will be  
399 critical to assessing these outcomes.

400 Increasing global temperatures could also affect the geographical distribution of the  
401 pathogen, driving a range shift to cooler higher latitude and higher elevation climates. Such a  
402 shift has been observed in another natural fungal disease, white pine blister rust. Dudney et al.  
403 (2021) found that, from 1995 to 2017, white pine blister rust prevalence has increased at high  
404 elevations and decreased at low elevations. A similar range shift to higher elevations could also  
405 happen in the anther-smut hosts for which the host range extends into higher elevations than the  
406 pathogen range. However, in *S. latifolia*, Bruns et al. (2018) found that anther smut is already  
407 present at the upper host range limits in the Italian Alps. Therefore, the ability of *Microbotryum*  
408 to shift its range in response to climate change could be limited by the range of the host.

409        The temperature sensitivity of *Microbotryum* could also lead to evolutionary changes in  
410    both the host and pathogen which could potentially feedback to affect disease prevalence  
411    (Jiranek et al. 2023). Specifically, the ability of high temperatures to prevent new infections and  
412    to temporarily cure existing infections could reduce selection for host resistance by allowing  
413    infected, susceptible genotypes to successfully reproduce during the hot summer months. Given  
414    that smut resistance can be costly (Biere and Antonovics 1996), this could result in selection  
415    favoring more susceptible genotypes in hot climates, which could in turn allow disease to persist  
416    even as the warming reduces the duration of transmission. Furthermore, the ability of the  
417    pathogen to adapt to warmer temperatures is currently unknown, but evidence of thermal  
418    adaptation has been observed in other fungal plant pathogens (Mboup et al. 2012, Boixel et al.  
419    2022).

420

#### 421    **Acknowledgements**

422    We thank Kaela Coil, Samuel Hulse, Eirena Li, Yanelyn Perez, and Alex Peska for help in the  
423    field and setting up the *in vitro* assays and Janis Antonovics for collecting the *Microbotryum*  
424    strains. We gratefully acknowledge greenhouse support from Meghan Fisher and Sydney  
425    Wallace and field support at the MAES Beltsville Station from Kevin Conner. The project was  
426    supported by the National Science Foundation (DEB-1936334 to EB). DC was supported by an  
427    NSF REU, Maryland Summer Scholars, and a UMD Honors college grant.

428

#### 429    **Conflict of interest statement**

430    We have no conflicts of interests to declare.

431

432 **Author contributions**

433 DC, SS, and EB designed the experiments. DC carried out experiments 1 and 2. AK carried out  
434 experiment 3. DC and SS analyzed the data. DC and EB wrote the manuscript, with input from  
435 all authors.

436

437 **References**

438 Abbate, J. L., and J. Antonovics. 2014. Elevational disease distribution in a natural plant-  
439 pathogen system: insights from changes across host populations and climate. *Oikos*  
440 123:1126–1136.

441 Alexander, H. M. 2010. Disease in natural plant populations, communities, and ecosystems:  
442 insights into ecological and evolutionary processes. *Plant Disease* 94:492–503.

443 Alexander, H. M., and J. Antonovics. 1988. Disease spread and population dynamics of anther-  
444 smut infection of *Silene alba* caused by the fungus *Ustilago violacea*. *The Journal of*  
445 *Ecology* 76:91.

446 Bebber, D. P., M. A. T. Ramotowski, and S. J. Gurr. 2013. Crop pests and pathogens move  
447 polewards in a warming world. *Nature Climate Change* 3:985–988.

448 Bernasconi, G., J. Antonovics, A. Biere, D. Charlesworth, L. F. Delph, D. Filatov, T. Giraud, M.  
449 E. Hood, G. A. B. Marais, D. McCauley, J. R. Pannell, J. A. Shykoff, B. Vyskot, L. M.  
450 Wolfe, and A. Widmer. 2009. *Silene* as a model system in ecology and evolution.  
451 *Heredity* 103:5–14.

452 Bever, J. D., S. A. Mangan, and H. M. Alexander. 2015. Maintenance of plant species diversity  
453 by pathogens. *Annual Review of Ecology, Evolution, and Systematics* 46:305–325.

454 Biere, A., and J. Antonovics. 1996. Sex-specific costs of resistance to the fungal pathogen

455 *Ustilago violacea* (*Microbotryum violaceum*) in *Silene alba*. *Evolution* 50:1098–1110.

456 Boixel, A., M. Chelle, and F. Suffert. 2022. Patterns of thermal adaptation in a globally  
457 distributed plant pathogen: Local diversity and plasticity reveal two-tier dynamics.

458 *Ecology and Evolution* 12.

459 Borer, E. T., P. R. Hosseini, E. W. Seabloom, and A. P. Dobson. 2007. Pathogen-induced  
460 reversal of native dominance in a grassland community. *Proceedings of the National  
461 Academy of Sciences* 104:5473–5478.

462 Bruns, E.L., J. Antonovics, C. Valentina, M. Hood. 2017. Transmission and temporal dynamics  
463 of anther-smut disease (*Microbotryum*) on alpine carnation (*Dianthus pavonius*). *Journal  
464 of Ecology* 105:1413-1424.

465 Bruns, E.L., J. Antonovics, and M.E. Hood. 2018. Is there a disease-free halo at species range  
466 limits? The codistribution of anther-smut disease and its host species. *Journal of Ecology*  
467 107:1–11.

468 Budria, A., and U. Candolin. 2014. How does human-induced environmental change influence  
469 host-parasite interactions? *Parasitology* 141:462–474.

470 Chaloner, T. M., S. J. Gurr, and D. P. Bebber. 2021. Plant pathogen infection risk tracks global  
471 crop yields under climate change. *Nature Climate Change* 11:710–715.

472 Cohen, J. M., M. D. Venesky, E. L. Sauer, D. J. Civitello, T. A. McMahon, E. A. Roznik, and J.  
473 R. Rohr. 2017. The thermal mismatch hypothesis explains host susceptibility to an  
474 emerging infectious disease. *Ecology Letters* 20:184–193.

475 Denchev, C. M., T. Giraud, and M. E. Hood. 2008. Three new species of anthericolous smut  
476 fungi on Caryophyllaceae. *Mycologia Balcanica* 6:79–84.

477 Dudney, J., C. E. Willing, A. J. Das, A. M. Latimer, J. C. B. Nesmith, and J. J. Battles. 2021.

478 Nonlinear shifts in infectious rust disease due to climate change. *Nature Communications*  
479 12:5102.

480 Elmqvist, T., D. Liu, U. Carlsson, and B. E. Giles. 1993. Anther-smut infection in *Silene dioica*:  
481 variation in floral morphology and patterns of spore deposition. *Oikos* 68:207.

482 Fontaine, M. C., P. Gladieux, M. E. Hood, and T. Giraud. 2013. History of the invasion of the  
483 anther smut pathogen on *Silene latifolia* in North America. *New Phytologist* 198:946–  
484 956.

485 Gilbert, G. S. 2002. Evolutionary ecology of plant diseases in natural ecosystems. *Annual Review*  
486 of *Phytopathology* 40:13–43.

487 Gilman, S. E., M. C. Urban, J. Tewksbury, G. W. Gilchrist, and R. D. Holt. 2010. A framework  
488 for community interactions under climate change. *Trends in Ecology & Evolution*  
489 25:325–331.

490 Giraud, T., R. Yockteng, M. López-Villavicencio, G. Refrégier, and M. E. Hood. 2008. Mating  
491 system of the anther smut fungus *Microbotryum violaceum*: Selfing under heterothallism.  
492 *Eukaryotic Cell* 7:765–775.

493 Gsell, A. S., A. Biere, W. de Boer, I. de Bruijn, G. Eichhorn, T. Frenken, S. Geisen, H. van der  
494 Jeugd, K. Mason-Jones, A. Meisner, M. P. Thakur, E. van Donk, M. P. Zwart, and D. B.  
495 Van de Waal. 2023. Environmental refuges from disease in host–parasite interactions  
496 under global change. *Ecology* 104:e4001.

497 Hood, M. E., J. I. Mena-Alí, A. K. Gibson, B. Oxelman, T. Giraud, R. Yockteng, M. T. K.  
498 Arroyo, F. Conti, A. B. Pedersen, P. Gladieux, and J. Antonovics. 2010. Distribution of  
499 the anther-smut pathogen *Microbotryum* on species of the Caryophyllaceae. *New*  
500 *Phytologist* 187:217–229.

501 IPCC, 2022: Climate Change 2022: Impacts, Adaptation, and Vulnerability. Contribution of  
502 Working Group II to the Sixth Assessment Report of the Intergovernmental Panel on  
503 Climate Change [H.-O. Pörtner, D.C. Roberts, M. Tignor, E.S. Poloczanska, K.  
504 Mintenbeck, A. Alegría, M. Craig, S. Langsdorf, S. Löschke, V. Möller, A. Okem, B.  
505 Rama (eds.)]. Cambridge University Press. Cambridge University Press, Cambridge, UK  
506 and New York, NY, USA, 3056

507 Jiranek, J., I. F. Miller, R. An, E. Bruns, and C. J. E. Metcalf. 2023. Mechanistic models to meet  
508 the challenge of climate change in plant–pathogen systems. *Philosophical Transactions  
509 of the Royal Society B: Biological Sciences* 378:20220017.

510 Kido, A., and M. E. Hood. 2020. Mining new sources of natural history observations for disease  
511 interactions. *American Journal of Botany* 107:3–11.

512 Liu, X., Z. Ma, M. W. Cadotte, F. Chen, J. He, and S. Zhou. 2019. Warming affects foliar fungal  
513 diseases more than precipitation in a Tibetan alpine meadow. *New Phytologist* 221:1574–  
514 1584.

515 Mboup, M., B. Bahri, M. Leconte, C. De Vallavieille-Pope, O. Kaltz, and J. Enjalbert. 2012.  
516 Genetic structure and local adaptation of European wheat yellow rust populations: the  
517 role of temperature-specific adaptation: Climate adaptation and genetic structure.  
518 *Evolutionary Applications* 5:341–352.

519 Mordecai, E. A. 2011. Pathogen impacts on plant communities: unifying theory, concepts, and  
520 empirical work. *Ecological Monographs* 81:429–441.

521 Mordecai, E. A., J. M. Cohen, M. V. Evans, P. Gudapati, L. R. Johnson, C. A. Lippi, K.  
522 Miazgowicz, C. C. Murdock, J. R. Rohr, S. J. Ryan, V. Savage, M. S. Shocket, A.  
523 Stewart Ibarra, M. B. Thomas, and D. P. Weikel. 2017. Detecting the impact of

524 temperature on transmission of Zika, dengue, and chikungunya using mechanistic  
525 models. *PLOS Neglected Tropical Diseases* 11:e0005568.

526 Mordecai, E. A., K. P. Paaijmans, L. R. Johnson, C. Balzer, T. Ben-Horin, E. de Moor, A.  
527 McNally, S. Pawar, S. J. Ryan, T. C. Smith, and K. D. Lafferty. 2013. Optimal  
528 temperature for malaria transmission is dramatically lower than previously predicted.  
529 *Ecology Letters* 16:22–30.

530 Nowakowski, A. J., S. M. Whitfield, E. A. Eskew, M. E. Thompson, J. P. Rose, B. L. Caraballo,  
531 J. L. Kerby, M. A. Donnelly, and B. D. Todd. 2016. Infection risk decreases with  
532 increasing mismatch in host and pathogen environmental tolerances. *Ecology Letters*  
533 19:1051–1061.

534 Penczykowski, R. M., E. Walker, S. Soubeyrand, and A. Laine. 2015. Linking winter conditions  
535 to regional disease dynamics in a wild plant–pathogen metapopulation. *New Phytologist*  
536 205:1142–1152.

537 Power, A. G., and C. E. Mitchell. 2004. Pathogen spillover in disease epidemics. *The American*  
538 *Naturalist* 164:S79–S89.

539 Raftoyannis, Y., and M. W. Dick. 2002. Effects of inoculum density, plant age and temperature  
540 on disease severity caused by pythiaceous fungi on several plants. *Phytoparasitica*  
541 30:67–76.

542 Ragonese, I.G., M.R. Sarkar, R.J. Hall, and S. Altizer. 2024. Extreme heat reduces host and  
543 parasite performance in a butterfly–parasite interaction. *Proc. R. Soc. B* 291: 20232305.

544 Robin, C., A. Andanson, G. Saint-Jean, O. Fabreguettes, and C. Dutech. 2017. What was old is  
545 new again: thermal adaptation within clonal lineages during range expansion in a fungal  
546 pathogen. *Molecular Ecology* 26:1952–1963.

547 Roche, B. M., H. M. Alexander, and A. D. Maltby. 1995. Dispersal and disease gradients of  
548 anther-smut infection of *Silene alba* at different life stages. *Ecology* 76:1863–1871.

549 Rossi, V., A. Ravanetti, E. Pattori, and S. Giosuè. 2001. Influence of temperature and humidity  
550 on the infection of wheat spikes by some fungi causing fusarium head blight. *Journal of*  
551 *Plant Pathology* 83(3):189-198.

552 Scala, E., M. Micheli, F. Ferretti, G. Maresi, F. Zottoli, B. Piškur, and L. Scattolin. 2019. New  
553 diseases due to indigenous fungi in a changing world: The case of hop hornbeam canker  
554 in the Italian Alps. *Forest Ecology and Management* 439:159–170.

555 Schäfer, A. M., M. Kemler, R. Bauer, and D. Begerow. 2010. The illustrated life cycle of  
556 *Microbotryum* on the host plant *Silene latifolia*. *Botany* 88:875–885.

557 Shah, D. A., and G. C. Bergstrom. (2000). Temperature dependent seed transmission of  
558 *Stagonospora nodorum* in wheat. *European Journal of Plant Pathology* 106:837-842.

559 Shocket, M. S., A. T. Strauss, J. L. Hite, M. Šljivar, D. J. Civitello, M. A. Duffy, C. E. Cáceres,  
560 and S. R. Hall. 2018. Temperature drives epidemics in a zooplankton-fungus disease  
561 system: A trait-driven approach points to transmission via host foraging. *The American*  
562 *Naturalist* 191:435–451.

563 Shocket, M. S., Verwillow, A. B., Numazu, M. G., Slamani, H., Cohen, J. M., El Moustaid, F.,  
564 Rohr, J., Johnson, L. R., & Mordecai, E. A. 2020. Transmission of West Nile and five  
565 other temperate mosquito-borne viruses peaks at temperatures between 23°C and 26°C.  
566 *eLife*, 9.

567 Stukenbrock, E., and S. Gurr. 2023. Address the growing urgency of fungal disease in crops.  
568 *Nature* 617:31–34.

569 Taylor, D. R., and S. R. Keller. 2007. Historical range expansion determines the phylogenetic

570 diversity introduced during contemporary species invasion: Historical effects on invasion  
571 genetics. *Evolution* 61:334–345.

572 Tracy, A. M., M. L. Pielmeier, R. M. Yoshioka, S. F. Heron, and C. D. Harvell. 2019. Increases  
573 and decreases in marine disease reports in an era of global change. *Proceedings of the*  
574 *Royal Society B: Biological Sciences* 286:20191718.

575 Ward, J., K. Kim, and C. Harvell. 2007. Temperature affects coral disease resistance and  
576 pathogen growth. *Marine Ecology Progress Series* 329:115–121.

577 Wood, S. N. (2006). Generalized Additive Models: An Introduction with R. Chapman and  
578 Hall/CRC.

579 Wood, S. N. 2011. Fast stable restricted maximum likelihood and marginal likelihood estimation  
580 of semiparametric generalized linear models: Estimation of semiparametric generalized  
581 linear models. *Journal of the Royal Statistical Society: Series B (Statistical Methodology)*  
582 73:3–36.

583 Zhan, J., L. Erickson, and J.J. Burton. 2018. Climate change accelerates local disease extinction  
584 rates in a long-term wild host–pathogen association. *Global Change Biology*, 24:3526–  
585 3536.

586

587 **Table 1.** The effect of temperature on *in vitro* life history traits of *M. lychnidis-dioicae* in  
 588 Experiment 1 and their predicted thermal thresholds. In all cases temperature was fit as a  
 589 smoothing parameter, in a GAM model. Germination rate was fit with a knots of 5, and colony  
 590 growth and conjugation rate was fit with a knots of 6 (See Appendix S1: Figure S1 for details).  
 591 Statistical tests are the result of a likelihood ratio test from a full model containing replicate and  
 592 parametric predictors of strain and mating type (See Appendix S1: Figure S1 for details).

| Life History<br>Trait | estimated<br>df | F      | p-value | Deviance<br>explained | Lower<br>threshold | Optimum | Upper<br>threshold |
|-----------------------|-----------------|--------|---------|-----------------------|--------------------|---------|--------------------|
| Germination           |                 |        |         |                       |                    |         |                    |
| rate                  | 3.658           | 34.49  | <0.0001 | 64.1%                 | 4.5 °C             | 13.5 °C | > 30 °C            |
| Colony Growth         | 4.915           | 270.90 | <0.0001 | 73.3%                 | 10.6 °C            | 20 °C   | 29.0 °C            |
| Conjugation rate      | 4.615           | 142.29 | <0.0001 | 75.2%                 | < 4 °C             | 10.5 °C | 25.7 °C            |

593

594 **Table 2:** Summary of a quasibinomial GLM showing the effect of temperature-treatment, plant  
595 family, and replicate on seedling infection rate in Experiment 2.

| Source                | Df | Deviance | Residual Df | Residual deviance | p       |
|-----------------------|----|----------|-------------|-------------------|---------|
| Temperature Treatment | 4  | 53.482   | 50          | 92.55             | <0.0001 |
| Plant Family          | 3  | 38.988   | 54          | 146.03            | <0.0001 |
| Replicate dish        | 2  | 16.327   | 57          | 177.42            | 0.0117  |

596

597 **Table 3:** Summary of a binomial GLM showing the effects of temperature, inoculation  
598 treatment, and plant family on seedling mortality in Experiment 2.

| Source                | Df | Deviance | Residual Df | Residual Deviance | p    |
|-----------------------|----|----------|-------------|-------------------|------|
| Temperature treatment | 4  | 6.91     | 71          | 79.46             | 0.14 |
| Inoculation treatment | 1  | 5.33     | 70          | 74.13             | 0.02 |
| Family                | 3  | 8.56     | 67          | 65.57             | 0.04 |

599

600 **Figure captions:**

601 **Figure 1:** Relationship between temperature and *in vitro* spore traits measured in Experiment 1.

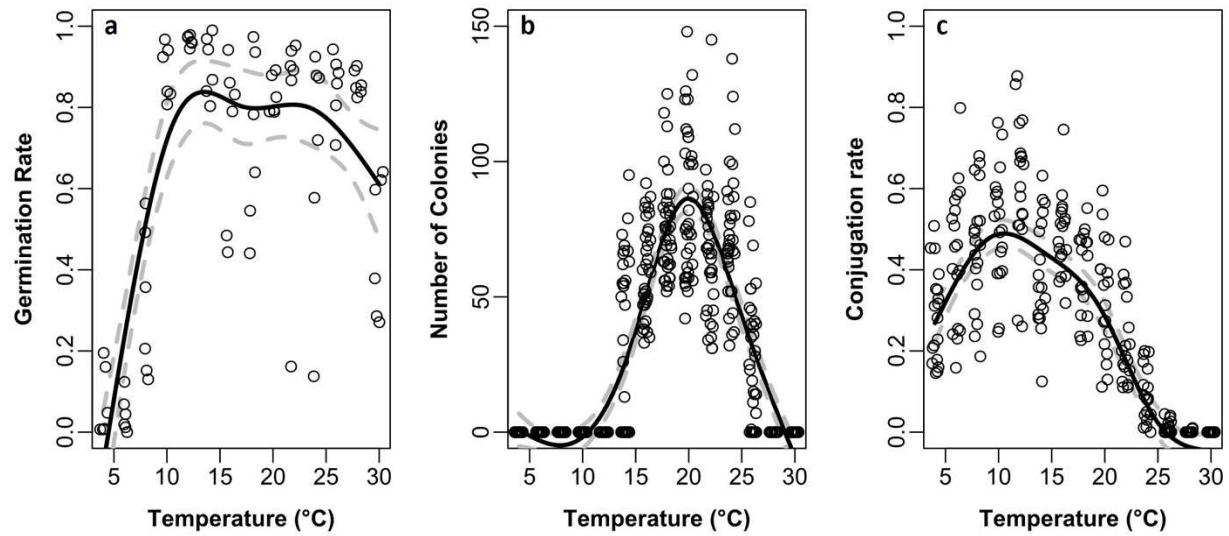
602 (a) Germination rate was calculated as a proportion of germinated to total teliospores. (b) Growth  
603 was measured by the number of visible colonies that grew after one week following plating of  
604 approximately 100 sporidia. (c) Conjugation rate was calculated as a proportion of conjugating  
605 sporidia to total sporidia. GAM models (black lines) are plotted with the 95% prediction  
606 confidence interval (gray dotted lines).

607

608 **Figure 2:** Proportion of seedlings that became infected with Microbotryum across different  
609 temperature treatments in Experiment 2. Each shape/color represents the proportion of plants  
610 diseased for each of the four plant families tested. Treatment means and standard errors are  
611 predicted from the glm model. Significance of each treatment relative to the control was  
612 determined by a Dunnett test and denoted \* $P < 0.05$  and \*\*\* $P < 0.001$  (See Appendix S1: Table  
613 S3).

614

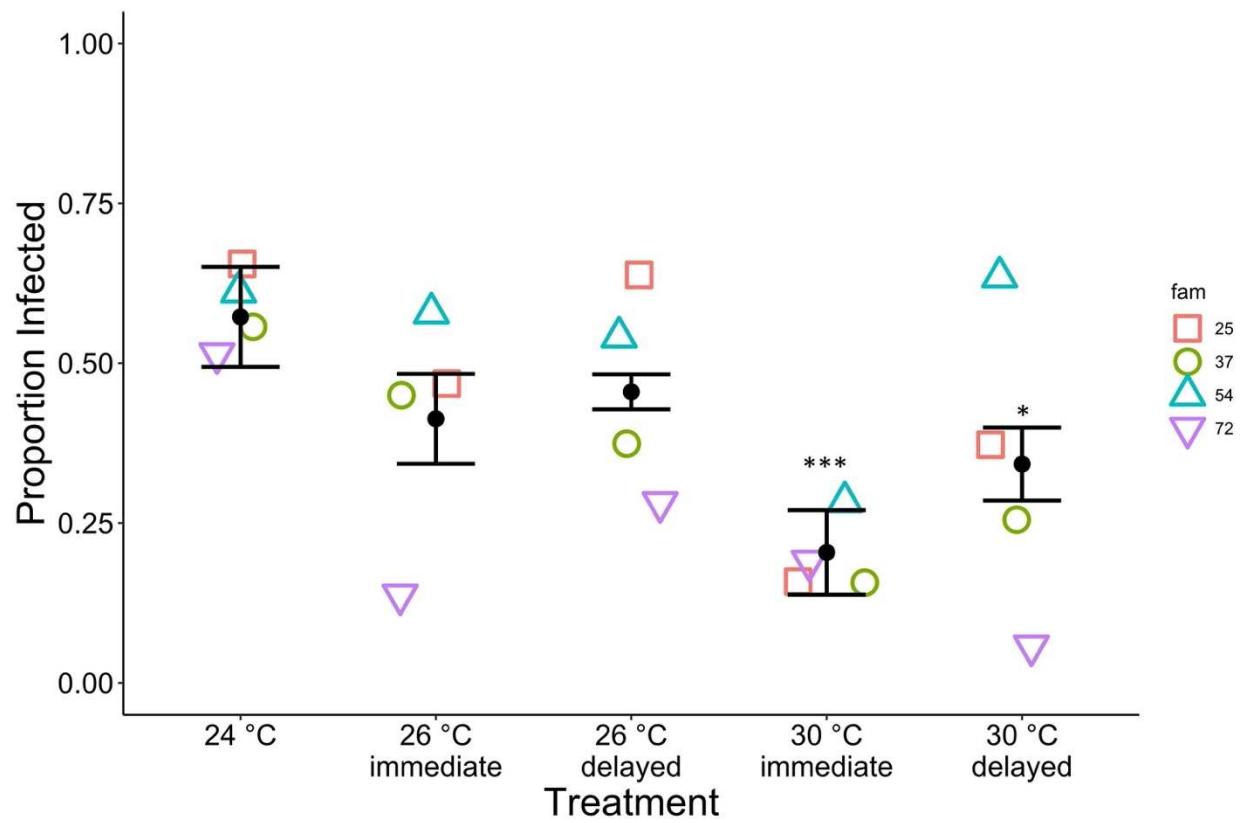
615 **Figure 3:** Flowering and disease expression rates of infected *Silene latifolia* plants in the field  
616 (Experiment 3). A) Daily high temperature at the field site in Beltsville, MD from May to  
617 November 2021. Gray lines show day to day fluctuation, red line is a smoothed spline fit. B)  
618 Total number of flowering fully healthy cured flowering (blue dots and smoothed line) and  
619 partially or fully symptomatic plants (black dots and lines). C-D) Relationship between the  
620 average high temperature of the preceding 14 days and C) total flowering proportion, and D)  
621 proportion of flowering plants expressing anther-smut symptoms. Error bars are 95% confidence  
622 prediction intervals.



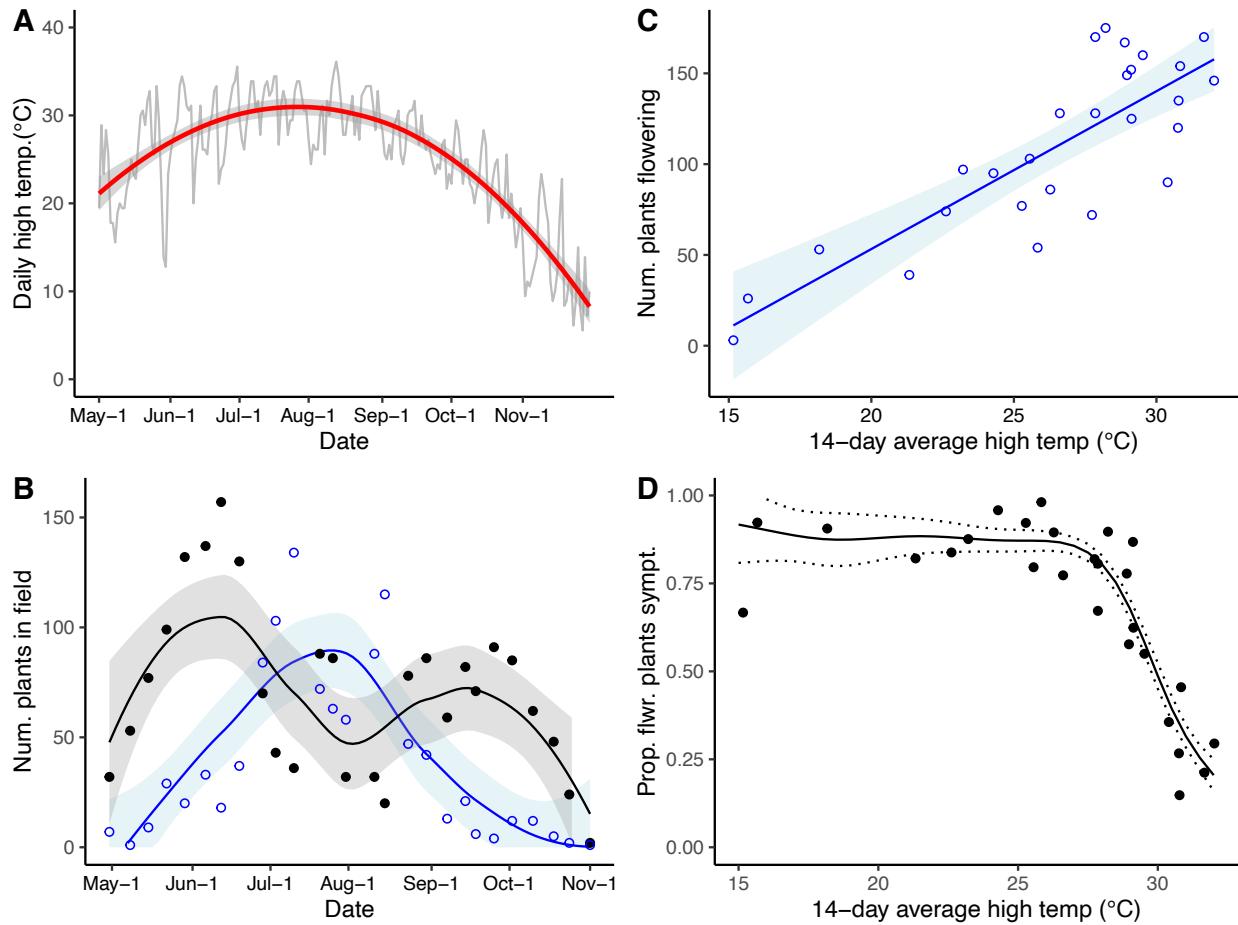
623

624

625



626



627

628

629 SUPPLEMENTAL MATERIAL: APPENDIX 1

630

631

632 **High temperatures reduce growth, infection, and transmission of a naturally occurring**  
633 **fungal plant pathogen**

634 By: Dalia V. Chen, Samuel P. Slowinski, Allyson K. Kido, Emily L. Bruns

635 Journal: Ecology- Article

636

637 **Supplemental methods:**

638 **Dilution plating method for generating single mating type sporidial colonies.**

639 Diploid teliospores were germinated for approximately 24 hours on 1% potato dextrose  
640 agar (PDA) plates until haploid sporidia formed. Plates were then washed with sterile DI water  
641 and dilution-plated onto a fresh PDA plate to generate single-sporidial colonies. Sporidial  
642 colonies were selected from each original teliospore culture and mating type was assessed  
643 through PCR with *M. lychnidis-dioicae* mating type specific primers (Xu et al. 2016). To amplify  
644 an a1 mating type specific sequence, the primers used were: MvSl\_phero\_a1\_F: 5'-AGCC  
645 TGTGCACCGGATAG-3' and MvSl\_phero\_a1\_R: 5'-ACACCTCCAGCCTCAATAC  
646 TAACATCTC-3'. To amplify an a2 specific sequence, the primers used were:  
647 MvSl\_phero\_a2\_F: 5'-AGCCGCCTCGAAGAGC-3' and MvSl\_phero\_a2\_R: 5'-  
648 AGTTCCGAAGGGCCACA-3' (Xu et al. 2016). PCR product was run on a 1% agarose gel, and  
649 mating type was determined based on the presence or absence of visible bands. Mating type was  
650 confirmed with conjugation assays on 1% water agar plates (*Microbotryum* sporidia of  
651 opposite mating type form visible conjugation tubes on low nutrient media).

652

653

654

655

656 **Table S1.** Full summary of best-fit GAM models for *in vitro* pathogen traits. There were no  
657 significant interaction effects between temperature and either strain or mating type.

#### Germination rate

| Term        | Fit Type       | df      | Resid.df | F     | p        |
|-------------|----------------|---------|----------|-------|----------|
| Temperature | Smooth, fixed  | 3.658   | 4.000    | 34.49 | < 0.0001 |
| Replicate   | Smooth, random | <0.0001 | 1.00     | 0     | 0.742    |

#### Colony growth

| Term        | Fit Type          | df     | Resid.df | F       | p        |
|-------------|-------------------|--------|----------|---------|----------|
| Temperature | Smooth, fixed     | 4.9170 | 5.0      | 276.883 | < 0.0001 |
| Replicate   | Smooth, random    | 0.7405 | 1.0      | 2.853   | 0.0502   |
| Mating type | Parametric, fixed | 1      | NA       | 1.095   | 0.296    |
| Strain      | Parametric, fixed | 2      | NA       | 6.418   | 0.0018   |

#### Conjugation rate

| Term        | Fit Type          | df     | Resid.df | F      | p        |
|-------------|-------------------|--------|----------|--------|----------|
| Temperature | Smooth, fixed     | 4.6198 | 5.0      | 144.25 | < 0.0001 |
| Replicate   | Smooth, random    | 0.9649 | 1.0      | 27.52  | < 0.0001 |
| Mating type | Parametric, fixed | 2      | NA       | 2.673  | 0.0711   |

658

659

660 **Table S2:** Summary statistics showing deviance explained and Un-biased Risk Estimator  
661 (UBRE) values for binomial gam models of proportion spore producing plants given different  
662 temperature time blocks. Models with the lower UBRE scores are better fits to the data (Wood  
663 2006). All models were fit with k=7, using package mgcv.

| Temperature   | Deviance     |               |
|---------------|--------------|---------------|
| period (days) | explained    | UBRE          |
| 7             | 67.8%        | 10.026        |
| 10            | 60.7%        | 12.244        |
| 12            | 73.2%        | 8.2168        |
| <b>14</b>     | <b>81.3%</b> | <b>5.5529</b> |
| 16            | 81.2%        | 5.5166        |

664  
665  
666  
667  
668  
669  
670  
671  
672

673 **Table S3:** A) Summary of one-way ANOVA showing the effect of temperature-treatment on  
674 proportion of seedlings infected in Experiment 2. Each temp treatment includes 3 replicates of 4  
675 host genotypes (n = 60 proportion values). B) results of a post-hoc Dunnett's test comparing to  
676 the control treatment.

**A**

| Source         | Sum Squares | Df | Mean Squares | F    | p value |
|----------------|-------------|----|--------------|------|---------|
| Temp Treatment | 1.00        | 4  | 0.25         | 6.97 | 0.00106 |
| Residuals      | 2.582       | 55 | 0.047        |      |         |

677

**B**

| Comparison                | Difference | Adjusted p-value |
|---------------------------|------------|------------------|
| Control vs 26°C immediate | -0.18      | 0.16             |
| Control vs 26°C delayed   | -0.13      | 0.42             |
| Control vs 30°C immediate | -0.39      | 0.00025          |
| Control vs 30°C delayed   | -0.25      | 0.02             |

678

679 **Table S4:** Summary of two-line linear regression approach with variable temperature cutoffs.

680 For each model we fit two separate linear regressions between 14-day average high temp and the  
681 proportion of flowering plants that show anther-smut symptoms. Overall residual error was  
682 minimized at 27°C.

| Cutoff temp (°C) | df1 | df2 | resid.error.1 | resid.error2 | total error      |
|------------------|-----|-----|---------------|--------------|------------------|
| 25               | 5   | 18  | 0.0942        | 0.1344       | 0.2286           |
| 26               | 8   | 15  | 0.08826       | 0.1315       | 0.21976          |
| 27               | 10  | 13  | 0.08888       | 0.1258       | <b>0.21468</b> * |
| 28               | 13  | 10  | 0.09704       | 0.1283       | 0.22534          |
| 29               | 16  | 7   | 0.1091        | 0.1423       | 0.2514           |

683

684

685

686

687

688

689

690

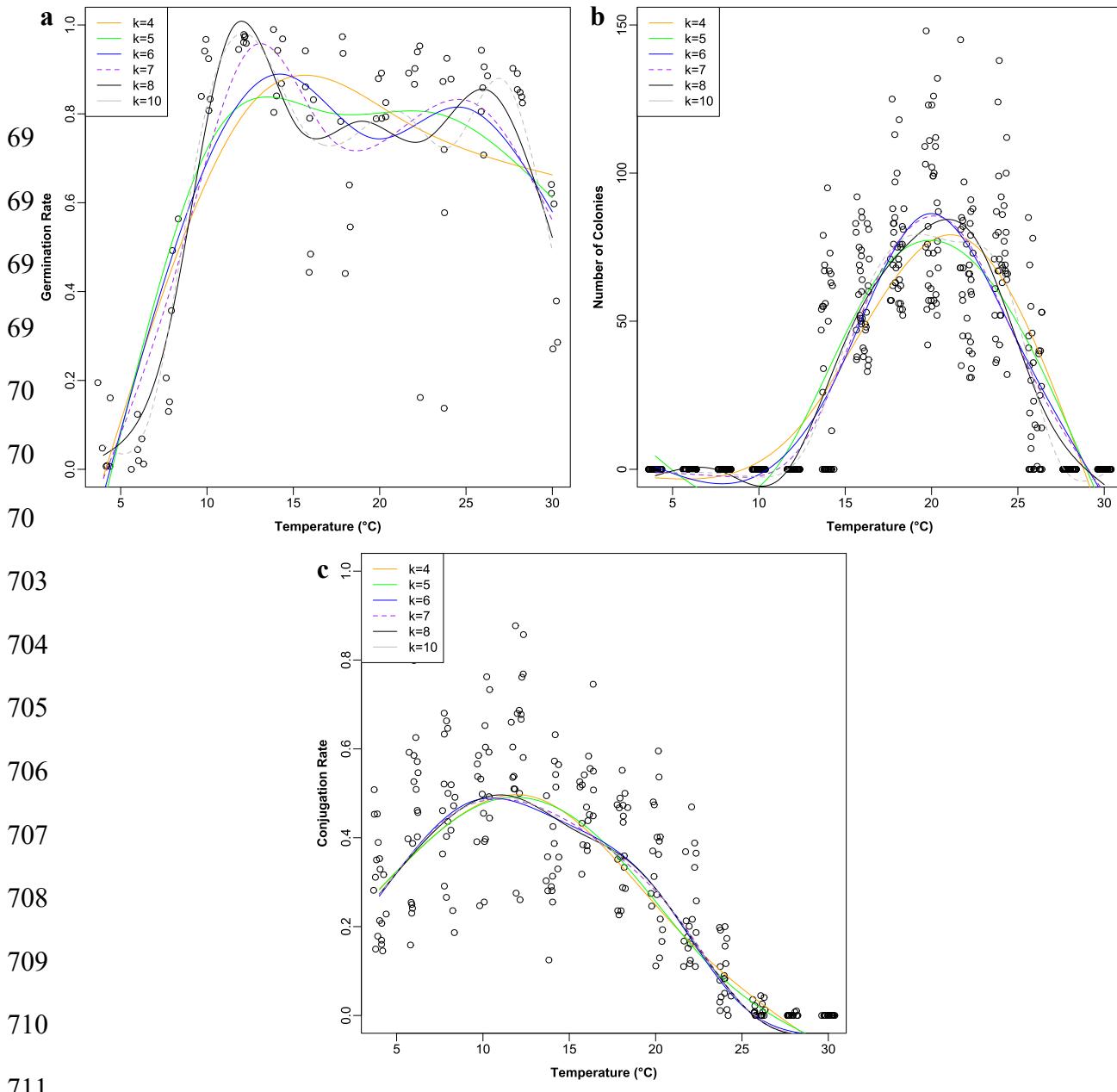
691

692

693

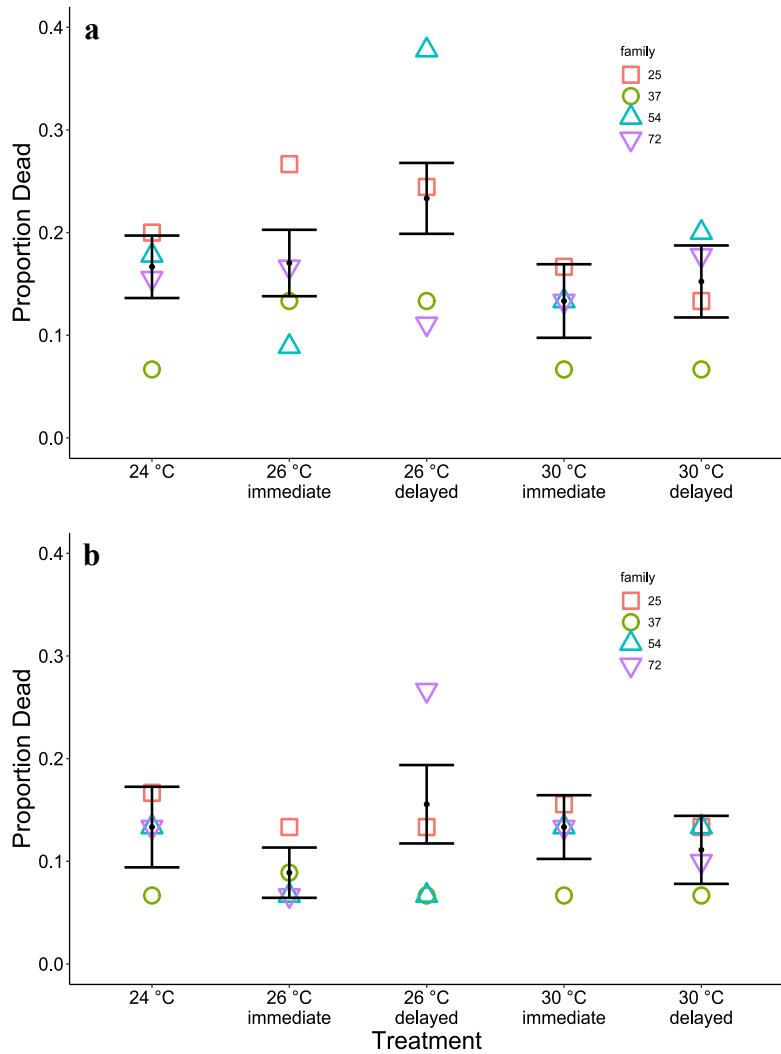
694

695



712 **Figure S1:** Generalized additive models with different number of knots fit to (a) germination  
 713 rate, (b) number of colonies, and (c) conjugation rate data over a range of temperatures in  
 714 Experiment 1. These graphs were used to choose the number of knots for the final model (k=5  
 715 for germination, k=6 for growth and conjugation). The black dots are the raw germination rate  
 716 data, and each line is a generalized additive model fit using a different number of knots.  
 717 Estimates of thermal maxima were generally robust to the choice of knots.

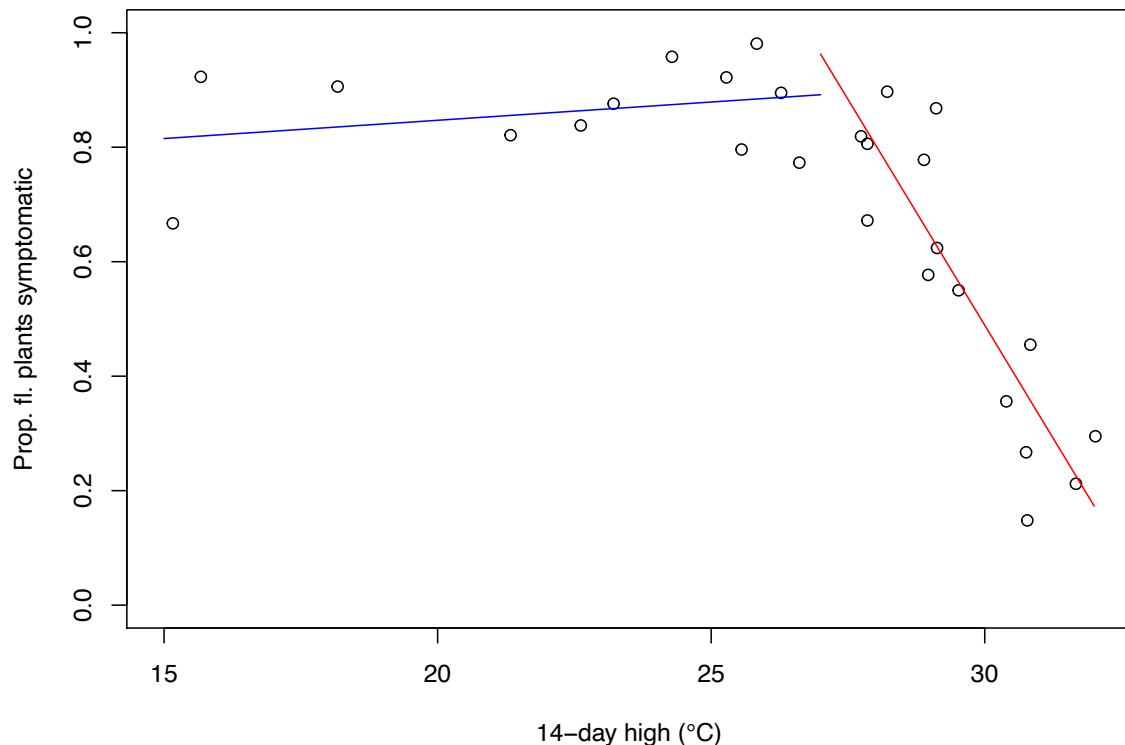
718



719

720

721 **Figure S2:** Proportion of dead seedlings from each temperature and time treatment in  
 722 Experiment 2. (a) inoculated with a pathogen solution. (b) mock-inoculated with water. Each  
 723 shape/color represents the proportion of plants dead for each of the four plant families tested.  
 724 The mean proportion is represented by the black dot. Error bars are the standard error of the  
 725 mean.



726

727 **Figure S3:** Best fit two-line model between average 14-day high temp and disease expression.

728 The cutoff temperature here is 27°C F. The slope of the first, blue line is not significantly

729 different from zero (slope=0.0064, df=10, se=0.0066, t=0.972 , p=0.3539), but the second red

730 line is highly significant (slope = -0.15788, df=13, se= 0.0238, t=-6.624, p<0.0001).

731

732

733 **References**

734 Wood, S. N. (2006). *Generalized Additive Models: An Introduction with R*. Chapman and  
735 Hall/CRC.

736 Xu, L., E. Petit, and M. E. Hood. 2016. Variation in mate-recognition pheromones of the fungal  
737 genus *Microbotryum*. *Heredity* 116:44–51.