Can hot temperatures limit disease transmission? A test of mechanisms in a zooplankton-

fungus system

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ABSTRACT

1. Thermal ecology theory predicts that transmission of infectious diseases should respond unimodally to temperature, i.e., be maximized at intermediate temperatures and constrained at extreme low and high temperatures. However, empirical evidence linking hot temperatures to decreased transmission in nature remains limited.

2. We tested the hypothesis that hot temperatures constrain transmission in a zooplankton-fungus (*Daphnia dentifera-Metschnikowia bicuspidata*) disease system where autumnal epidemics typically start after lakes cool from their peak summer temperatures. This pattern suggested that maximally hot summer temperatures could be inhibiting disease spread.

3. Using a series of lab experiments, we examined the effects of high temperatures on five mechanistic components of transmission. We found that (1) high temperatures increased exposure to parasites by speeding up foraging rate but (2) did not alter infection success post-exposure. (3) High temperatures lowered parasite production (due to faster host death and an inferred delay in parasite growth). (4) Parasites made in hot conditions were less infectious to the next host (instilling a parasite 'rearing' or 'trans-host' effect of temperature during the prior infection). (5) High temperatures in the free-living stage also reduce parasite infectivity, either by killing or harming parasites.

4. We then assembled the five mechanisms into an index of disease spread. The resulting unimodal thermal response was most strongly driven by the rearing effect. Transmission peaked at intermediate-hot temperatures (25-26°C) and then decreased at maximally hot temperatures (30-32°C). However, transmission at these maximally hot temperatures only trended slightly lower than the baseline control (20°C), which easily sustains epidemics

in laboratory conditions and in nature. Overall, we conclude that while exposure to hot

epilimnetic temperatures does somewhat constrain disease, we lack evidence that this

effect fully explains the lack of summer epidemics in this natural system. This work

demonstrates the importance of experimentally testing hypothesized mechanisms of

thermal constraints on disease transmission. Furthermore, it cautions against drawing

conclusions based on field patterns and theory alone.

INTRODUCTION

How do high temperatures affect the spread of infectious diseases? In the current prevailing view, warming from climate change will shift the geographic range of diseases: some new areas will become warm enough to support disease, whereas others that previously sustained disease will become too hot (Altizer, Ostfeld, Johnson, Kutz, & Harvell, 2013; Lafferty, 2009; Lafferty & Mordecai, 2016). This hypothesis stems from a principle of thermal biology: most biological traits have unimodal reaction norms, where performance peaks at intermediate temperatures and declines to zero at cooler and warmer temperatures (Dell, Pawar, & Savage, 2011). Thus, once temperatures exceed the thermal optima of traits driving transmission, disease should decline. Many models predict upper thermal constraints on diseases, such as for helminthic parasites of Arctic ungulates (Molnár, Kutz, Hoar, & Dobson, 2013), a rhizocephalan crab parasite (Gehman, Hall, & Byers, 2018), human schistosomiasis (Mangal, Paterson, & Fenton, 2008), a microsporidian *Daphnia* parasite (Kirk et al., 2018), and mosquito-borne diseases (Mordecai et al., 2017, 2013; Paull et al., 2017; Shocket, Ryan, & Mordecai, 2018; Tesla et al., 2018). Additionally, there is evidence for upper thermal constraints on disease in natural populations of grasshoppers infected by a fungal pathogen (Carruthers, Larkin, & Firstencel, 1992), the rhizocephalan crab parasite (Gehman et al., 2018), amphibians infected by chytrid fungus (Berger et al., 2004; Raffel, Michel, Sites, & Rohr, 2010), bats infected with white-nose fungus (Langwig et al., 2015), and mosquito-borne human pathogens (Gatton, Kay, & Ryan, 2005; Mordecai et al., 2013; Peña-García, Triana-Chávez, & Arboleda-Sánchez, 2017; Perkins, Metcalf, Grenfell, & Tatem, 2015). However, temperature often co-varies with other seasonal environmental factors (Altizer et al., 2006), so causally linking temperature to observed field patterns of disease is challenging (Altizer et al., 2006; Pascual & Dobson, 2005). Thus, the

generality of upper thermal constraints excluding disease remains unclear.

Conceptually, upper thermal constraints act like fever, taking advantage of a common

thermal mismatch between hosts and parasites. Because hosts can often withstand hotter

environments than their parasites, many animals increase their body temperature when infected

(Thomas & Blanford, 2003; see others below). In endothermic vertebrates, fever arises internally

via cytokines and the central nervous system (Conti, Tabarean, Andrei, & Bartfai, 2004). In

ectotherms, fever arises from behavioral thermoregulation (microhabitat selection) and is

widespread, occurring in amphibians (Richards-Zawacki, 2009; but see Sauer et al., 2018),

reptiles (Zimmerman, Vogel, & Bowden, 2010), fish (Boltaña et al., 2013), snails (Zbikowska,

Wrotek, Cichy, & Kozak, 2013), and insects (Anderson, Blanford, Jenkins, & Thomas, 2013;

Carruthers et al., 1992; McClain, Magnuson, & Warner, 1988; Stahlschmidt & Adamo, 2013;

Thomas & Blanford, 2003). In both types of fever, high temperatures can impair parasite

performance, enhancing clearance or reducing virulence of infection. An analogous process can

occur within ectothermic hosts inhabiting high ambient temperatures (regardless of infection

status)—in essence, an environmental fever. High ambient temperatures can also harm parasites

with free-living stages outside of hosts. Mechanistically linking high temperatures to reduced

disease requires examining thermal effects on components of the transmission process

(McCallum et al. 2017). We use the term 'transmission (process)' to broadly refer to the full

parasite life cycle, including infective propagule production and propagule survival in the

environment; we also use 'transmission rate' as it is narrowly defined as the rate of new

infections (i.e., the parameter 'β' calculated from the infection prevalence and the density of

hosts and parasites; McCallum et al. 2017).

Here, we evaluate several possible mechanisms for upper thermal constraints on disease

transmission in a planktonic-fungal system. Autumnal epidemics start once lake waters cool

below summer maxima (Fig. 1A). These delayed starts could reflect hot temperatures inhibiting

disease if they push any of five transmission components past their thermal optima. First, hot

temperatures could slow host feeding and lower consumption-based exposure to parasites.

Second, hot temperatures could lower parasite infectivity inside hosts, lowering the probability

of successful infection (via effects on hosts and/or parasites). Third, hot temperatures could

decrease the quantity of parasite propagules [spores] produced by an infection. This decrease

could stem from slower host growth rate (since parasite production often scales with host

growth: Hall, Knight, et al., 2009; Hall, Simonis, Nisbet, Tessier, & Cáceres, 2009), slower

parasite growth independent from host growth, or enhanced mortality of infected hosts

(truncating production time; Auld, Hall, Housley Ochs, Sebastian, & Duffy, 2014; Civitello,

Forys, Johnson, & Hall, 2012). Fourth, hot temperatures could lower the quality of parasite

spores released from dead hosts into the environment (Shocket, Vergara, et al., 2018). Finally,

these free-living spores could be harmed or killed by hot temperatures. Thus, high temperatures

could constrain this fungal disease at multiple stages of the transmission process.

Using a series of experiments, we tested upper thermal constraints on transmission via the

five mechanisms described above (Fig. 1B). Hot temperatures increased exposure to parasites by

speeding up foraging rate (f, mechanism 1) without lowering parasite infectivity within hosts (u,

mechanism 2). Hot temperatures reduced the number of parasites produced (σ, mechanism 3).

This decrease stemmed from faster host death and an inferred delay in parasite production, but

not depressed host or parasite growth rates. Parasites produced at hot temperatures were less

infectious to the next host, instilling a parasite 'rearing' or 'trans-host' effect on spore quality (ρ,

mechanism 4), analogous to trans-generational effects in hosts (Shocket, Vergara, et al., 2018).

Hot temperatures also either harmed or killed parasites in the free-living stage, lowering their

estimated infectivity (φ, mechanism 5). We assembled these effects into an index of potential

disease spread ('transmission potential'). The resulting unimodal thermal response was driven

most strongly by the rearing effect. Potential transmission peaked at intermediate-hot

temperatures (25-26°C) and then decreased at maximally hot temperatures (30-32°C). However,

potential transmission at maximally hot temperatures only trended slightly lower than the

baseline control temperature (20°C), which easily sustains epidemics in laboratory conditions

and nature. Overall, we conclude that exposure to high temperatures does somewhat constrain

disease, but we lack evidence that it fully explains the lack of summer epidemics in this system.

132   **STUDY SYSTEM**

The hosts (*Daphnia dentifera*; hereafter 'hosts') are dominant zooplankton grazers in

many freshwater temperate lakes across the Midwestern United States (Tessier & Woodruff,

2002). Some populations experience epidemics of the virulent fungal parasite *Metschnikowia*

biscupidata (hereafter 'fungus'; Hall et al., 2010; Penczykowski, Hall, Civitello, & Duffy, 2014).

Maximum prevalence can reach up to 60% (Penczykowski, Hall, et al., 2014). Hosts become

infected when they filter-feed on algae and inadvertently consume fungal spores (Hall et al.,

2007). The spores pierce the host's gut wall, entering its body cavity. Inside, fungal conidia

replicate in the hemolymph before maturing into new spores (Stewart Merrill & Cáceres, 2018).

Following host death from infection, spores are released into the water for new hosts to consume

(Ebert, 2005).

The seasonality of epidemics motivated the present focus on high temperatures.

Epidemics typically begin in late summer or early fall (August–October) and wane in late fall or

early winter (November–December; Hall, Becker, Duffy, & Cáceres, 2011; Penczykowski, Hall,

et al., 2014). During this time, lake water temperature declines (Shocket, Strauss, et al., 2018).

Many traits that influence disease spread (demographic traits of hosts, transmission rate, and

spore production) change plastically with temperature (Hall, Tessier, Duffy, Huebner, &

Cáceres, 2006; Shocket, Strauss, et al., 2018). Transmission increases with constant temperatures

up to 26°C, and hosts cannot be cultured in constant temperatures above 27°C (Shocket, Strauss,

et al., 2018). However, organisms can withstand otherwise lethal high temperatures in a

fluctuating environment (Niehaus, Angilletta, Sears, Franklin, & Wilson, 2012). For instance, in

our stratified lakes, hosts regularly experience temperatures exceeding 27°C (typically maxima

of 29-32°C; Fig. 1A) in summer: they migrate between the colder, deeper hypolimnion during

day (to avoid mortality from visually oriented fish predators) and the warmer, upper epilimnion

at night (to take advantage of greater algal resources and faster growth in warmer temperatures)

(Hall, Duffy, Tessier, & Cáceres, 2005; Lampert, 1989). Epidemics often begin as lakes start to

cool from maximum summer temperatures (Fig. 1A). This pattern suggested that high

temperatures could constrain disease spread, as predicted by theory of thermal disease ecology

(Lafferty, 2009; Lafferty & Mordecai, 2016).

METHODS

Field Survey

Field survey data generated the motivating pattern, i.e. the relationship between epidemic

start date and epilimnetic temperature (Fig. 1A). We surveyed 10–28 lakes in southwestern

Indiana (Greene and Sullivan Counties) weekly (2009 – 2011) or bi-weekly (2013 - 2014) from

August to December. For each visit, we collected a zooplankton sample (13 cm diameter net

with 153 μm mesh) and measured lake water temperature data at 0.5–1 meter intervals with a

Hydrolab multiprobe (Hach Environmental). For each sample, we visually diagnosed 400+ live

hosts with a dissecting scope (20-50X magnification). An epidemic 'started' when infection

prevalence first exceeded 1% for two consecutive sampling visits (Shocket, Strauss, et al., 2018).

We calculated the epilimnetic temperature by fitting a spline to temperature across water depth,

and averaging from the water surface to the depth where the temperature gradient first exceeded

$1°C\ m^{-1}$ (i.e., the thermocline; see Hite et al. 2016 Appendix S2).

General Approach

We measured how high temperatures influence five mechanistic components of the

transmission process with laboratory assays (Table 1). Then we combined them into a synthetic

index of disease spread: 'transmission potential' (Auld et al., 2014). For mechanisms involving

the host or host-parasite interaction (mechanisms 1-3: foraging rate [f], spore infectivity from

within-host processes [u], and spore yield [σ]), we used fluctuating temperatures to expose hosts

to high temperatures for part of the day (they cannot survive constant temperatures >27°C).

Hosts were kept on a 16:8 light:dark cycle. All hosts experienced the same 20°C temperature for

8 hours, and then 20, 26, or 32°C for 16 hours ('maximum temperature'). For mechanisms 4-5

(rearing effect on spore quality [ρ] and free-living spore effect [φ]), we conducted common

garden infection assays (using uniform groups of hosts at constant 20°C) with spores from

different treatments. Thus, variation in transmission rate can be attributed to differences in spore

infectivity. Temperatures varied slightly among experiments (25 or 26°C, 30 or 32°C) based on

incubator availability. For calculating transmission potential, we treat temperature categorically

and pool these treatments.

For all parameters, we bootstrapped 95% confidence intervals (data sampled within

groups, with replacement; 10,000 samples). For parameters calculated at the population level (β,

u, ρ, and φ), we used randomization tests to compare temperature treatments (treatment labels

shuffled among host individuals, without replacement; 10,000 samples). For σ we used model

selection with AIC to compare treatments. For f and transmission potential (for which traditional

statistical tests were not available), we used the bootstrapped distributions to compare treatments.

Specifically, we calculated the cumulative probability density of the best estimate from one

treatment according to the bootstrapped distribution of the other. These PD-values are analogous

to p-values. We considered treatments significantly different if $PD<0.025$. See Appendix for

details and a complete list of statistical tests and results.

Due to time and incubator constraints, we were unable to replicate the experiment across

multiple incubators. Thus, our temperature treatments are 'pseudo-replicated' in that all the

replicates for a single temperature treatment were conducted in the same incubator at the same

time. Accordingly, our results may be influenced by random incubator effects.

Mechanisms 1 & 2: Foraging rate (f) and spore infectivity from within-host processes (u)

We measured foraging rate of hosts by comparing the fluorescence of ungrazed and

grazed algae (Penczykowski, Lemanski, et al., 2014; Sarnelle & Wilson, 2008). We added

estimates of foraging rate at 30°C to those at 20 and 25°C presented elsewhere using the same

methods (Shocket, Vergara, et al., 2018). In both experiments, we measured foraging rate across

a gradient of host body size (Kooijman, 2009) to index foraging at a common size among

experiments (1.5 mm). We used maximum likelihood estimation (MLE) to fit size-dependent

functions of foraging with the 'bbmle' package (Bolker & R Development Core Team, 2017) in

R (R Core Team, 2017). See Appendix for details.

We measured how high temperature impacts transmission rate (β) and spore infectivity

from within-host processes (u) with an infection assay ('$\beta + u$ measurement assay'). For

successful infection, the fungus must break through the host gut barrier and then develop and

replicate within the host hemolymph. High temperatures could inhibit the parasite during either

process. Thus, we factorially manipulated the maximum temperature (20 and 32°C) during

parasite exposure and infection establishment (for four exposure/infection establishment

treatments: 20/20, 20/32, 32/20, and 32/32°C). This approach reveals if high temperature

interferes with initial gut penetration or subsequent establishment of infection (similar to Allen &

Little, 2011). Hosts were exposed individually in their 'exposure temperature' for 24 hours, then

moved to their 'infection establishment temperature.' Later, hosts were visually diagnosed for

infection. Transmission rate was calculated from proportion infected (see Appendix). We

calculated spore infectivity from within-host processes (u) for each treatment by dividing

transmission rate (β) by foraging rate (f) at the exposure temperature ($u=\beta/f$).

Mechanism 3: Spore yield (σ) and related host and parasite traits

We measured how high temperatures impact final spore yield (σ) of infected hosts that

died from their infection. This trait estimates release of spores into the environment. We pooled

spore yields from the $\beta + u$ measurement assay (above, treatments: 20/20 and 32/32°C) and the

within-host parasite growth assay (below, treatments = 20, 26, and 32°C) since they did not

differ statistically (20°C: $p = 0.65$; 32°C $p = 0.93$). We tested for differences between

temperatures by fitting a suite of models via MLE: in each model spore yield was normally

distributed and temperature treatments could exhibit the same or different means and standard

deviations. We compared models using AIC and calculated *p*-values with likelihood ratio tests.

To distinguish between three possible mechanisms driving the thermal response of spore

yield, we quantified related host and parasite traits. First, we measured host growth rate (g_h) with

a juvenile growth rate assay (Lampert & Trubetskova, 1996)(see Appendix), since spore yield

often scales with g_h (e.g., with different host food resources: Hall, Knight, et al., 2009; Hall,

Simonis, et al., 2009). We compared treatments with t-tests. Second, we measured parasite

growth (i.e., number of mature spores within hosts over time) using a sacrifice series ('within-

host parasite growth assay;' see Appendix), since spore yield could decline if the number of

parasites grows more slowly, independently of host condition (Thomas & Blanford, 2003). We

fit and bootstrapped linear models of 'spore load' over time to estimate parasite growth rate (g_p,

the model slope). 'Spore load' estimates included spores in living (i.e., sacrificed) hosts, unlike

'spore yield,' which was calculated only from dead hosts that were killed by the parasite. Spore

yield is directly relevant for the epidemiology of the system, while spore load simply measures

one of the underlying processes (parasite growth rate per day, g_p) that contributes to spore yield.

Spore load increased linearly over the full time series at 26 and 32ºC. Spore load plateaued after

day 19 at 20ºC, so we truncated the time series to estimate the linear slope for only that portion.

Finally, we calculated death rate (*d*) of the infected hosts used to estimate spore yield (see

Appendix), since spore yield can decline with shorter host lifespan (Auld et al., 2014; Civitello et

al., 2012). We compared treatments with randomization tests.

Mechanisms 4 & 5: rearing (ρ) and free-living spore (φ) effects on infectivity

We measured how high temperatures modify spore infectivity prior to encountering the

focal host via a rearing effect on baseline spore quality (ρ) and harm to free-living spores (φ). We

conducted infection assays on 'common garden' groups of hosts at 20°C using different spore

treatments (i.e., from different spore rearing temperatures for ρ and from different spore

incubation temperatures for φ). Thus, variation in transmission rate reflects differences in spore

infectivity. To measure ρ, we conducted two assays, one with spores produced in the $\beta + u$

measurement assay (20/20 and 32/32°C treatments only) and another with spores produced in the

within-host parasite growth assay (20, 26, and 32°C treatments). To measure φ, we used spores

incubated at three temperatures (20, 25, and 30°C) for two durations (1-day and 7-days) in

constant, non-fluctuating temperatures (since spores do not migrate between stratified water

layers). One-day incubations were stored at 4°C for the first six days (standard procedure for

spore storage). For both experiments, we estimated transmission rates (β) from the prevalence

data (see Appendix).

Both mechanisms influence transmission by modifying spore infectivity (already

estimated from within-host processes as u, mechanism 1). Thus, in order to incorporate these

additional mechanisms into a synthetic metric for disease spread (transmission potential, see

below), we calculated unit-less rearing (ρ) and free-living (φ) effects standardized to infectivity

at 20°C. Specifically, we calculated the parameters by dividing the estimates for transmission

rate (β) at 26 and 32°C by that at 20°C. Given this calculation, values of $\rho < 1$ or $\varphi < 1$ mean

spores are less infectious due to rearing or free-living effects than at 20°C, respectively;

conversely, values > 1 mean spores are more infectious than at 20°C. To calculate confidence

intervals at 20°C, we divided the bootstrapped distribution by a randomly-shuffled version of

itself. Additionally, because harm to free-living spores occurs over time as spores are removed

by hosts, we used a simple model to estimate time-weighted transmission rates for φ. We

assumed that spore infectivity declined linearly over the 7-day assay (above), and that hosts

consume spores at a constant foraging rate (resulting in an exponential decay in spores remaining

over time). Thus, we weighted the estimated transmission rate on each day by the proportion of

spores consumed by hosts on that day (see Appendix for detailed methods and a sensitivity

analysis of the time-weighted transmission model).

Transmission potential

We calculated an index of disease spread to synthesize the effects of all five mechanisms.

We defined transmission potential as the product of all five parameters ($f\,u\,\sigma\,\rho\,\varphi$). We generated

confidence intervals using the bootstrapped parameter distributions. To visualize the contribution

of each parameter, we calculated transmission potential from all possible four-parameter

combinations, holding the fifth constant at its 20°C point estimate. These values reveal how each

parameter affects the magnitude and uncertainty of transmission potential (hence, it is a type of

sensitivity analysis).

297   **RESULTS**

Mechanisms 1 & 2: Foraging rate (f) and spore infectivity from within-host processes (u)

Contrary to our predictions, high temperature did not lower transmission rate (β) during

either stage (exposure or infection establishment; Fig. 2A). Instead, transmission rate was higher

when hosts were exposed at 32°C than at 20°C (20°C infection establishment: $p=0.0013$; 32°C

infection establishment: $p<0.0001$). Temperature during infection establishment exerted no effect

on transmission rate (20°C exposure: $p=0.10$; 32°C exposure: $p=0.31$). When exposure and

establishment temperatures were equal (as in nature; the 20/20 and 32/32°C treatments here),

transmission rate was higher at 32°C than at 20°C ($p=0.0068$). Thus, even at maximal

epilimnetic temperatures, the impacts of higher temperatures on transmission rate promoted

rather than inhibited disease.

The thermal response of transmission rate was mechanistically driven by foraging rate of

hosts (f), not spore infectivity from within-host processes (u). Foraging rate increased from 20 to

25°C ($PD=0$; see Methods and Appendix for a description of PD values, which are analogous but

not identical to p-values) and then plateaued at 30°C ($PD=0.11$; Fig. 2B). Thus, hosts encounter

more spores at 25 and 30°C than at 20°C. After we accounted for predicted host-parasite contact,

spore infectivity was fairly insensitive to high temperatures (Fig. 2C). Temperature during

infection establishment did not impact spore infectivity (20°C exposure: $p=0.10$; 32°C exposure:

$p=0.31$). Exposure temperature increased spore infectivity (20°C infection establishment:

$p=0.034$; 32°C infection establishment: $p=0.0052$), but in the opposite direction of the

hypothesized mechanism (hotter temperature increased infectivity). When exposure and infection

establishment temperatures were equal (as in nature), spore infectivity did not differ ($p=0.37$).

Thus, high temperatures increased the foraging rate of hosts, elevating host contact with spores,

while spore infectivity barely changed. These changes in parasite exposure led to more

transmission at high temperatures.

Mechanism 3: Spore yield (σ) and other measures of host and parasite growth

Final spore yield (σ) in hosts that died from infection was lower at 32°C than at 20 and

26°C (Fig. 3A; best-fitting model had two means, see table S2 for model AIC scores and Akaike

weights.) This pattern was not explained by host condition estimated via growth rate. Host

growth rate (g_h, Fig. 3B) always increased with temperature (20 versus 26°C: $p=4.7 \times 10^{-6}$; 26 vs.

32°C: $p=0.00038$). Instead, the pattern was explained by a combination of death rates of hosts

and delays in spore maturation. Infected hosts died more quickly at 26°C than at 20°C

($p<0.0001$), and death rate trended higher from 26 to 32°C ($p=0.063$; Fig. 3C). Meanwhile,

growth rate of mature parasite spores (g_p, time series in Fig. 3D, linear slopes [growth rate] in

Fig. 3E) did not change with temperature (all $PD>0.15$). However, temperature did affect the

timing of initial spore production within hosts (i.e., intercepts of linear model). At the earliest

point in the sacrifice series (day 8), spore load was highest at 26°C, intermediate at 32°C, and

nearly zero at 20°C (Fig. 3D). Given thermally insensitive daily growth rates of parasites (g_P;

Fig. 3E), these head-starts were maintained over time (Fig. 3D). This effect on early spore

production, coupled with host death rate (Fig 3C), explains the spore yield pattern. Final spore

yield was lower at 32 than 26°C because there were fewer spores initially (on day 8) and hosts

died more quickly (less time to produce spores). At 20°C, spore production started even later, but

the delay was compensated for by the much longer lifespans of infected hosts (lower death rate,

d; Fig. 3C).

Mechanisms 4 & 5: rearing (ρ) and free-living spore (φ) effects on infectivity

Spore infectivity (measured as transmission rate) responded unimodally to temperature in

the previous infection (rearing effect on spore quality, ρ; Fig. 4A). Infectivity increased

significantly for spores made at 20 versus 26°C for one of two spore sources ($p=0.0083$ for

spores from $\beta + u$ measurement assay [square, Fig 4A]; $p=0.092$ for spores from within-host

growth assay [diamond]). Infectivity then declined for spores made at 26 versus 32°C ($p=0.0001$

for both spore sources). Infectivity was significantly lower for spores made at 32 versus 20°C for

one of two spore sources ($p=0.16$ for spores from $\beta + u$ measurement assay [square]; $p=0.026$ for

spores from within-host growth assay [diamond]). The parameter ρ (Fig. 4C) shows the rearing

effect pooled for both spore sources and normalized by transmission rate at 20°C (for use in the

transmission potential calculation).

The thermal environment of free-living spores also impacted their infectivity (φ; Fig

4B,D). Spore infectivity decreased with higher incubation temperatures after 7 days (20 versus

25°C: p=0.0031; 25 versus 30°C: p<0.0001; diamonds on Fig. 4B). However, spore infectivity

did not change after 1-day incubations (flat line in Fig. 4B [squares]; 20 versus 25°C: p=0.65, 25

versus 30°C: p=0.64). All 1-day incubations used stored (refrigerated) spores. They had lower

infectivity than the 7-day incubation at 20°C, likely because storage at 4°C also lowers spore

infectivity (1 versus 7-day incubations at 20°C: p<0.0001; Duffy *unpublished data*). The

parameter φ (Fig. 4D) shows the free-living spore effect assuming that spores lose infectivity

gradually over seven days as they are consumed by hosts (see *Methods* and Appendix) and

normalized by transmission rate at 20°C (for use in the transmission potential calculation).

Transmission potential ($f\,u\,\sigma\,\rho\,\varphi$)

Transmission potential, the product of all mechanisms ($f\,u\,\sigma\,\rho\,\varphi$), responded unimodally

to high temperatures. This metric first increased from 20 to 25/26°C (PD=0.017); then, it

decreased from 25/26 to 30/32°C (PD=0.0001; 'full transmission potential'; Fig. 5A).

Transmission potential at 30/32°C trended lower than at 20°C (PD=0.11), but this difference was

not significant. Thus, high temperatures do not constrain disease enough via these five

mechanisms to explain absence of summer epidemics.

The initial increase in transmission potential from 20 to 25/26°C was driven most

strongly by host foraging (f, mechanism 1) and the rearing effect on spore quality (ρ, mechanism

4): holding either trait constant removes the significant difference between temperatures (Fig. 5B

and 5E, respectively). The subsequent drop in transmission potential from 25/26 to 30/32°C was

driven most strongly by the rearing effect (ρ): holding it constant again removes the significant

difference (Fig. 5E). Harm to free-living spores (φ, mechanism 5) also contributes somewhat

(Fig. 5F vs. Fig. 5A), though not enough to affect the statistical significance. Additionally, the

thermal response of host foraging (f) is key for maintaining transmission at high temperatures:

without increased exposure to spores, the remaining mechanisms would significantly reduce

transmission at 30/32°C compared to 20°C (Fig 5B). Spore infectivity from within-host

processes (u, mechanism 2) and spore yield (σ, mechanism 3) had no effect (Fig. 5C vs. Fig. 5A)

and very little effect (Fig. 5D vs. Fig. 5A) on transmission potential, respectively.

385   **DISCUSSION**

We investigated upper thermal constraints on fungal epidemics in a *Daphnia* zooplankton

host. The seasonality of the autumnal epidemics suggested that hot conditions might constrain

disease: epidemics usually start after lakes cool from maximal summer temperatures in the

epilimnion (29-32°C). We tested five potential thermal constraints on transmission. First,

foraging (exposure) rate of hosts (f) increased at high temperatures (Fig 2B), while, second, high

temperatures did not affect the infectivity of spores from within-host processes (u; Fig 2C). Thus,

high temperatures increased transmission rate, β (where $\beta=uf$; Fig 2A). Third, spore yield (σ)

declined slightly at 32°C (Fig 3A). Fourth, a rearing effect on spore quality driven by

temperature during the previous infection (ρ) emerged: spores made at 32°C were less infectious

than those made at 26°C and sometimes 20°C (for one of two spore sources, Fig 4A). Finally,

harm to free-living spores (φ) lowered infectivity as temperature increased (Fig 4B). Overall,

transmission potential is much lower at 32°C than at 26°C, but still similar to at 20°C (Fig 5A), a

398 temperature that easily supports epidemics in both nature (Shocket, Strauss, et al., 2018) and

399 laboratory environments (Civitello et al., 2012; Shocket, Strauss, et al., 2018). Thus, it seems that

400 maximally high temperatures do constrain disease, but not sufficiently to explain the absence of

401 summer epidemics on their own.

402  Contrary to our initial hypothesis, high temperatures *increased* transmission rate (Fig

403 2A). In principle, high temperatures can lower infection success if pathogens tolerate heat less

404 well than hosts (Thomas & Blanford, 2003). For instance, many ectothermic hosts behaviorally

405 induce fever to reduce the negative costs of infection (Anderson et al., 2013; Carruthers et al.,

406 1992; Elliot, Blanford, & Thomas, 2002; Thomas & Blanford, 2003). Further, fungi are

407 particularly sensitive to high temperatures compared to other pathogen taxa (Robert &

408 Casadevall, 2009) and fungal pathogens are often limited by temperature (Berger et al., 2004;

409 Carruthers et al., 1992; Langwig et al., 2015; Raffel et al., 2010; Thomas & Blanford, 2003).

410 However, high temperatures did not interfere with this fungus's success at either stage of

411 transmission: the day of exposure, when most spores penetrate the host's gut, or infection

412 establishment, when the fungus develops and replicates within the host (Stewart Merrill &

413 Cáceres, 2018). Instead, high temperatures elevated host foraging rate (Fig 2B), which increases

414 exposure to parasites, thereby increasing transmission rate (Hall et al., 2007). In lakes, the

415 thermal response of foraging (exposure) drives variation in the size of epidemics, which occur in

416 autumn: epidemics that start earlier in warmer conditions grow larger than those starting later

417 and colder (Shocket, Strauss, et al., 2018). This foraging-controlled exposure to parasites is a

418 potentially general mechanism: higher temperatures increase outbreak size for armyworms that

419 consume virus particles on leaves (Elderd & Reilly, 2014). However, transmission plateaued

420 with temperature for another ingested *Daphnia* pathogen (Vale, Stjernman, & Little, 2008).

Spore yield (σ) declined at the highest temperature (32°C; Fig. 3). Although the effect on

transmission potential was minimal (Fig 5D), the results for related traits provide mechanistic

insights into host-parasite interactions. Parasite burdens often decline at temperatures near the

thermal maxima of the host and/or parasite, e.g., for nematodes in slugs (Wilson, Digweed,

Brown, Ivanonva, & Hapca, 2015), trematodes in snails (Paull, Lafonte, & Johnson, 2012),

bacteria in *Daphnia* (Vale et al., 2008) and fruit flies (Lazzaro, Flores, Lorigan, & Yourth, 2008),

and powdery mildew in plants (Laine, 2007). In theory, reduced parasite production at hot

temperatures could arise from several mechanisms. First, parasite production could decline if

host growth slows, since spore yield often scales with host growth, at least along resource

gradients (Hall, Knight, et al., 2009; Hall, Simonis, et al., 2009). However, here host growth rate

(g_h) increased with temperature while spore yield was flat and then decreased (Fig 3B).

Therefore, spore production was decoupled from host growth rate (i.e., the link based on

resources did not apply for temperature). Second, the parasite itself could grow more slowly at

high temperatures. For example, high temperatures slow bacterial growth inside fruit flies

(Lazzaro et al., 2008), fungal growth in grasshoppers (Springate & Thomas, 2005), and fungal

growth on warm-adapted (but not cold-adapted) amphibians (Cohen et al., 2017). In contrast,

here parasite growth rate (g_p) did not respond to temperature (slope in Fig 3D; Fig 3E).

Instead, the decline in spore production at high temperatures arose from a combination of

host death rate and the timing of initial spore production. Temperature determined spore load on

day 8 (the earliest sampling time in the assay; Fig. 3D). Based on that information (and the

constant parasite growth rates, Fig 3E), we infer that spore production began earliest at 26°C,

followed by 32°C, and then 20°C. These head starts were maintained over time and explain the

spore yield pattern when combined with death rate of infected hosts (Fig. 3C). In general, shorter

lifespan of infected hosts decreases time for spore production, thereby depressing spore yield

(Auld et al., 2014; Civitello et al., 2012). Here, spore yield was lower at 32 than 26°C because

spore production started later and hosts died more quickly. At 20°C, spore production started

even later, but longer host lifespan compensated for this delay (i.e., the fungus had longer to

grow within hosts). Do similar patterns exist in other systems? Unfortunately, few studies focus

on traits underlying thermal responses of parasite load. Hence questions remain: How often does

temperature change the timing versus the rate of parasite production? How often does

temperature decouple positive relationships between host growth and parasite production? The

answers matter because spore yield can influence epidemic size for obligate killer parasites (like

the fungus here: Civitello et al., 2015). Thus, developing a general framework from data across

host-parasite systems remains a fruitful area for future research.

High temperatures reduced transmission potential via two effects on spore infectivity that

act outside the focal host. First, a rearing effect on spore quality (ρ) driven by temperature of

spore production in the previous host elevated (26°C) and then lowered (32°C) spore infectivity

(compared to 20°C). Rearing effects on parasite performance can arise with variation in

resources consumed by hosts (Cornet, Bichet, Larcombe, Faivre, & Sorci, 2014; Little, Birch,

Vale, & Tseng, 2007; Tseng, 2006), temperature experienced by hosts (Shocket, Vergara, et al.,

2018), or host genotype (Searle et al., 2015). These understudied rearing effects may drive

performance of parasites to an unappreciated extent (Shocket, Vergara, et al., 2018). Second,

harm to free-living spores (φ, including spore mortality) also inhibited infection at high

temperatures. After seven days in 30°C, spores lost 92% of their initial infectivity. This

constraint may arise in other systems: for example, high temperatures elevate mortality in free-

living helminths of Arctic ungulates (Molnár et al., 2013). However, in the planktonic system

here, 7-day result likely exaggerates the thermal constraint. While difficult to quantify, physical

sinking, consumption (Civitello, Pearsall, Duffy, & Hall, 2013; Penczykowski, Hall, et al., 2014;

Shocket, Vergara, et al., 2018; Strauss, Civitello, Cáceres, & Hall, 2015) and damage from

radiation (Overholt et al., 2012) likely remove most spores before 7 days. To acknowledge this

mortality, we weighted this component of infectivity (φ) using a model of spore longevity.

Assuming this modeled weighting reflects reality in lakes, this free-living effect lacks enough

strength to inhibit epidemics during summer, even when combined with the other mechanisms

(see Appendix for sensitivity analysis of the time-weighting model). However, more realistic

dynamical models and better resolved trait data for the free-living spore effect could change the

estimates for how high temperatures affect transmission.

477   Although the impact of temperature on these five mechanisms alone does not explain the

lack of epidemics during summer, other co-varying environmental factors could combine with

thermal effects to sufficiently inhibit transmission. Such factors include damage to free-living

spores by solar radiation (Overholt et al., 2012), consumption of spores by resistant zooplankton

species that are more abundant earlier in the year (Penczykowski, Hall, et al., 2014), and low

spore production due to poor quality of host food resources (Hall, Knight, et al., 2009). All of

these mechanisms could contribute to the observed field pattern, and interact with the thermal

effects examined here. Furthermore, covariation among drivers could be disrupted by climate

change. For example, high temperatures may persist later in the year when damaging solar

radiation is less intense. Incorporating the impacts of these other environmental factors may help

explain the current field pattern and improve predictions for the impact of climate change on

epidemics. However, any predictions for climate change and disease will need to explicitly

account for the effects of temperature variation and extremes.

Thermal variation and extremes have important and distinct impacts on organismal

performance. Variation changes trait values compared to constant temperatures with the same

mean (Dowd, King, & Denny, 2015). In most cases, thermal variation can be reasonably

incorporated using nonlinear averaging based on Jensen's Inequality ('JI'; Bernhardt, Sunda,

Thompson, & O'Connor, 2018; Ruel & Ayres, 1999). JI predicts that variation should generally

increase traits at cooler temperatures (below a trait optimum) and decrease them at higher

temperatures (above a trait optimum), and is supported empirically (Bernhardt et al., 2018;

Paaijmans et al., 2010, 2013). Beyond JI, extreme temperatures can cause additional effects via

acute thermal stress (Dowd et al., 2015). For instance, limits of species ranges may depend more

on tolerance of episodic thermal extremes than on the thermal responses of life history traits

(Overgaard, Kearney, & Hoffmann, 2014). Our study did not include perspectives based on JI or

extremes. We employed a relevant form of thermal variation, mimicking migratory behavior of

hosts in stratified lakes. However, we did not intend to isolate effects of thermal variation or to

expose hosts to thermal extremes. Future efforts could estimate both to better predict the impact

of climate change on the host, the parasite, and their interaction.

The current prevailing view argues that hot temperatures should constrain disease

transmission in nature (Altizer et al., 2013; Lafferty, 2009; Lafferty & Mordecai, 2016). This

constraint arises when unimodal thermal reaction norms depress key traits that drive disease

spread. However, such constraints have been rigorously tested in only a handful of systems.

Here, we hypothesized that high summer temperatures limit transmission of a zooplankton-

fungus disease system with autumnal epidemics (i.e., during cooler conditions). High

temperatures constrained disease transmission enough to produce a unimodal thermal response.

This response arose primarily through a rearing effect on spore quality and due to harm to free-

living spores. However, the thermal mechanisms estimated here were not sufficient explain the

lack of summer epidemics. Hence, we draw two major lessons. First, we need to continue to

rigorously evaluate multiple mechanisms of thermal constraints on components of disease

transmission. Second, our example cautions against drawing conclusions about constraints on

disease from warming based on field patterns and theory alone.

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Table 1: The experiments (and spore sources) used to test the five mechanistic components of

disease transmission.

Mechanism	Experiment(s)	Spore source(s)
1. Foraging rate (f, Fig. 2B)	Foraging rate assay	NA
2. Within-host spore infectivity (u, Fig. 2C)	$\beta + u$ measurement assay Foraging rate assay	General lab stock
3. Spore yield (σ, Fig. 3A)	$\beta + u$ measurement assay Within-host parasite growth assay	General lab stock
4. Rearing effect on infectivity (ρ, Fig 4A,C)	Common garden infection assay #1	$\beta + u$ measurement assay Within-host parasite growth assay
5. Free-living spore effect on infectivity (φ, Fig 4B,D)	Common garden infection assay #2	General lab stock incubated at different temperatures

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Figure 1: Motivating field pattern and mechanistic components of transmission. (A) Fungal
epidemics usually start (dark grey bars) after lakes have cooled from the maximum summer
temperature (light grey bars). Epidemics never started when the epilimnion (upper, warmer
layer) was hotter than 30°C, suggesting an upper thermal constraint. Data summarize 74
epidemics from 20 lakes in Indiana (USA) sampled from 2009-2015. (B) High temperature could
limit transmission via five mechanisms. 1-2) Hosts become infected at transmission rate β, which
can be divided into 1) host foraging rate (f), i.e., exposure to spores, and 2) spore infectivity, as
determined by within-host processes (u). 3) Parasite spores are produced at spore yield (σ). 4) A
rearing effect from temperature during the previous infection (ρ) determines initial spore
infectivity. 5) Harm to free-living spores (φ) might also impact their infectivity. The product of
all five components ($f\ u\ \sigma\ \rho\ \varphi$) determines 'transmission potential'.

Figure 2: High temperature impacts on transmission rate (*β*), foraging rate (*f*, mechanism 1) and spore infectivity from current within-host processes (*u*, mechanism 2). In A and C, the effect of high temperature during parasite exposure and infection establishment (20°C infection establishment = white circles, solid line; 32°C infection establishment = dark grey circles, dotted line). (A) Transmission rate (*β*) increased when hosts were exposed at 32°C and did not change with infection establishment temperature. For constant temperatures, transmission is higher at 32°C than at 20°C. (B) Foraging (exposure) rate of hosts (*f*) is higher at 26°C (light grey) and 32°C (dark grey) than at 20°C (white). (C) Spore infectivity (*u=β/f*) increased when hosts were exposed at 32°C for both infection establishment temperatures. However, for constant temperatures, infectivity did not differ between 20 and 32°C. Error bars show 95% CIs. Letters indicate significant differences.

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Figure 3: High temperature impacts on spore yield (σ, mechanism 3) and possible
underlying traits. (A) Final spore yield at host death (σ) was lower at 32°C (dark grey) than at
20°C (white) or 26°C (light grey). (B) Host growth rate (g_h) increased with temperature for all
treatments. (C) Death rate of infected hosts (d) increased from 20 to 26°C and trended higher
from 26 to 32°C. (D) Spore load within hosts through time at 32°C (dotted line), 26°C (dashed
line), and 20°C (solid line), fit with linear models. 26°C points are shifted over for visual clarity
of error bars. (E) Parasite growth rate (g_p, slopes of lines in panel D) did not change with
temperature. Hence, declining σ stems from higher death rate of infected hosts and low initial
parasite growth, not slower growth rates of hosts or parasites. (A-C,E) Error bars show 95% CIs.
(D) Error bars show SE; square points are single hosts. Letters indicate significant differences.

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Figure 4: High temperature impacts on a rearing effect (_ρ_, mechanism 4) and harm to free-
living spores (_φ_, mechanism 5). Variation in transmission rate from common garden infection
assays reflects differences in spore infectivity. (A) Spores came from the $\beta + u$ measurement
assay (Fig. 2; squares) and the within host parasite growth assay ('WHPG'; Fig. 3; diamonds).
Spore infectivity increased with rearing temperature from 20°C (white) to 26°C (light grey; $\beta + u$
only) and decreased with rearing temperatures from 26°C to 32°C (dark grey, both spore
sources). Spore infectivity was lower at 32°C than at 20°C (WHPG spores only). (B) Spore
infectivity decreased when free-living spores were incubated in high temperatures for 7 days but
not for 1 day. Storage at 4°C for 6 days (for all 1-day incubations) also lowered spore infectivity
relative to the 7-day incubation at 20°C. (C-D) Parameter values (transmission rate scaled by
values at 20°C) for (C) rearing effect, _ρ,_ and (D) free-living effect, _φ_. Phi values also based on
time-weighted model (see text for details). Error bars show 95% CIs. Letters indicate significant
differences.

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Figure 5: High temperature impacts on transmission potential. (A) Transmission potential
(*fuσρφ*) responds unimodally, increasing from 20°C (white) to 25/26°C (light grey) and
decreasing from 25/26°C to 30/32°C (dark grey). (B-F) Transmission potential with each
mechanism held constant to show sensitivity to each parameter: (B) foraging rate (*f*), (C) spore
infectivity from within-host effects (*u*), (D) spore yield (*σ*), (E) rearing effect (*ρ*), and (F) harm
free-living spores (*φ*). The rearing effect (E) has the largest impact on transmission potential
(hence, without it, the response of transmission potential is flat with temperature). Error bars
show 95% CIs. Letters indicate significant differences. Y-axis is ln-transformed.

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SUPPORTING INFORMATION

Additional supporting information may be found in the online version of this article.

Appendix S1: Methods, Figures, and Tables

Figure S1: Components of a simple model used to estimate parameter φ.

Figure S2: Sensitivity analysis for spore consumption model parameter (c) affecting damage to free-living spores (φ) and transmission potential.

Table S1: p-values from randomization tests.

Table S2: p-values and ΔAIC from model selection.

Table S3: PD (probability density) values for traits.

Table S4: PD (probability density) values for transmission potential 'sensitivity analysis' calculations.