

Physiological Costs of Altered Hydrothermal Conditions in Harvested Forests

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ABSTRACT

Amphibians are among the most rapidly declining vertebrate groups worldwide, with habitat modification and climate change driving widespread population declines. Much work on amphibian vulnerability focuses on gradual shifts of climate change. While in many landscapes, habitat-altering disturbances such as clearcutting introduce sudden shifts in thermal and hydric environments on timescales far shorter than those of potential local adaptation and potentially removing critical microhabitat refugia. Such changes may be particularly relevant for amphibians, whose behaviour and ecological performance is linked to both body temperature and dehydration status. However, predicting the impacts of harvesting on amphibian hydrothermal physiology is not straightforward as higher temperatures (up to thermal optima) may increase performance, while dehydration will decrease it. To untangle these effects, I compared hydrothermal performance curves for *Dryophytes versicolor* with water loss rates and operative temperatures of replica frogs in harvested, edge, and unharvested boreal forest across four time-since-cut stages. Replica frog water loss was significantly correlated with real frog water loss ($R^2 = 0.86$). I found performance declines past 15-20% dehydration and that baseline performance was lower at cooler temperatures and higher dehydrations. Clearcut environments reduced performance for gray treefrogs during overnight activity periods, particularly in uncovered microhabitats, across all time-since-cut groups, indicating that the increased hydrothermal vulnerability from harvesting is maintained through succession. By examining the relationship between performance, hydration, and temperature, we can begin to understand how removal or alteration of key microhabitats may impact individual fitness and population persistence within disturbed landscapes.

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This research has benefited from my presence on these lands, which are deeply shaped by ongoing histories of colonization. I am grateful and honoured to have had the opportunity to learn and work here, and I recognize that gratitude is not a substitute for action. Land acknowledgement is only a small part of building and maintaining meaningful relationships with the First Peoples of Canada. I am committed to continuing the work of learning from, supporting, and fostering equitable collaborations with Indigenous communities whose lands and knowledge systems continue to be vital to this work.

With this acknowledgement, I seek to honour the Indigenous peoples of the many places where I conducted fieldwork.

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1 INTRODUCTION

Amphibians are among the most rapidly declining vertebrate groups worldwide, with habitat modification and climate change driving widespread population declines (Hof et al., 2011; Luedtke et al., 2023; Stuart et al., 2004). Most work on anuran climate sensitivity assumes gradual change under shifting climate regimes (see Scheffers et al., 2013). However, in many landscapes, habitat-altering disturbances such as clearcutting introduce sudden and often extreme shifts in thermal and hydric environments (Franklin et al., 2002; Gardner et al., 2007; R. D. Semlitsch et al., 2009) on timescales far shorter than those of potential local adaptation (De Frenne et al., 2019), and potentially removing critical microhabitat refugia (Tuff et al., 2016). These rapid, localized, changes can fundamentally alter habitat quality by degrading or removing the microhabitats used for thermoregulation, hydration, and refuge.

Among the most critical factors influencing amphibian survival and performance are temperature and water availability, which together shape their metabolic processes, behavior, and overall fitness (Anderson & Andrade, 2017; Greenberg & Palen, 2021; Lertzman-Lepofsky et al., 2020; Rozen-Rechels et al., 2019). Ectothermic organisms, such as amphibians, are especially vulnerable to shifting climate because they rely on external environmental conditions to regulate their body temperature and hydration and maintain physiological function (Hillman et al., 2008; Huey et al., 2012; Wells, 2010). While the effects of temperature on amphibian function have been well studied (Cossins, 2012; Deutsch, 2008; Huey & Kingsolver, 1989; Pottier et al., 2025; Rollins-Smith & Le Sage, 2023; Sunday et al., 2014), dehydration also imposes considerable physiological costs, influencing locomotor capacity, metabolic efficiency, predator escape, and foraging success (Anderson & Andrade, 2017; Gatten, 1987; Rogowitz et al., 1999; Shoemaker et al., 1992).

To navigate the potential challenges of extreme hydrothermal habitat conditions, many ectotherms depend on specific microhabitats—such as shaded refuges, moist substrates, or hydrothermally preferable retreats—that offer opportunities for thermoregulation and hydration (Klinges et al., 2024). Microclimatic refuges may thus be essential for daily and seasonal survival and for long-term persistence. However, the availability and quality of these spaces are increasingly threatened by the rapid pace of habitat disturbance, land-use change, and the accelerating impacts of climate change, including rising temperatures and intensified drought (Luedtke et al., 2023; Zhang et al., 2021). Moreover, long-term dependence on retreats under increased thermal and hydric vulnerability comes with a cost as it can reduce available activity hours (Huey & Stevenson, 1979). Reduced activity time constrains essential behaviors such as foraging, mating, and dispersal, ultimately limiting energy acquisition and reproductive output (Rittenhouse et al., 2008; Spotila et al., 1992). Over time, these limitations can lead to reduced recruitment and population growth, and reliance on retreat sites has been shown to predict population-level declines under warming scenarios (Duarte et al., 2012; Sinervo et al., 2010).

While the effects of both temperature and hydration have been considered separately, few studies have jointly considered temperature and hydration in the context of real-world environmental conditions, despite evidence that these factors interact to shape behavior, distribution, and fitness (Anderson & Andrade, 2017; Greenberg & Palen, 2021; Lertzman-Lepofsky et al., 2020; Rozen-Rechels et al., 2019; Shoemaker & Nagy, 1977). Further, many microclimate studies rely on sampling at a single location and point in time or presence/absence data (Janin et al., 2011; Robinson et al., 2023) which fail to capture the effects of prolonged exposure or the consequences of chronic thermal stress, especially under fluctuating environmental conditions. Moreover, most eco-physiological models are diurnally biased,

underrepresenting the nocturnal activity patterns of amphibians (Kearney & Porter, 2004; Kearney & Predavec, 2000; Klinges et al., 2024; Rutschmann et al., 2024) and potentially resulting in misleading conclusions about species abilities to cope with environmental change.

In this thesis, I assess how rapid changes in microhabitats post-clearcutting affect the physiological capacities that are attainable for Eastern Gray Treefrogs (*Dryophytes versicolor*, Hylidae) within regenerating habitats. By examining these conditions across different stages of forest succession, I estimate how habitat quality shifts throughout 20-years of forest regrowth post-clearcut and evaluate the potential for forest-dwelling amphibians to persist under altered hydrothermal regimes.

1.1 Background

Forest harvesting reconfigures the ecological structure and function of North American forest ecosystems (Weber & Flannigan, 1997; Wolf et al., 2021). Many plant species possess adaptive mechanisms (e.g. shade tolerant birch and aspen) either to survive or to quickly recolonise cleared areas, and differences in recolonisation regimes amongst species post-disturbance results in a diversity of vegetational composition trajectories (Dawe et al., 2022; Grandpré et al., 1993). Over the past 50 years, forest harvesting has rapidly altered the structure and composition of the boreal forest primarily through removal of canopy, mid story, and understory vegetation cover (Bergeron & Fenton, 2012; Dupuch & Fortin, 2013; Keenan & Kimmins, 1993). Combined with favourable temperature conditions due to global climate warming, clearcutting regimes are expected to shift Boreal Forest composition towards forests dominated by broadleaf deciduous tree species, which regenerate quickly in full sunlight (Anyomi et al., 2022; Carleton & Maclellan, 1994; Dawe et al., 2022; Larocque et al., 2024).

Forest harvesting has both positive and negative effects on forest dwelling species, depending partially on vegetational composition during post-cut forest succession (Harper et al., 2015; Hocking & Semlitsch, 2008; Popescu & Hunter, 2011; Semlitsch et al., 2009; Todd et al., 2014). The effects of harvesting vary depending on pre-harvest stand composition and topography, the extent of canopy, debris, and vegetation removal, post-harvest management, and pre- and post-harvest climate and topography (Keenan & Kimmins, 1993; Carleton & Maclellan, 1994). More intensive cutting regimes that undergo multiple disturbance are more prone to overall forest composition shifts than single less intensive disturbances (Larocque et al., 2024). Forestry companies in Northwestern Ontario currently implement removal strategies that (cl)aim to simulate the effects of natural disturbances on forest composition (Natural Resources & Forestry, 2001). The dominant approach in this region is intensive even-aged management which leaves some retention trees or small patches of forest upon the landscape (Natural Resources & Forestry, 2023). However, clearcut cycles are relatively short compared to cycles of natural disturbance (Jõgiste et al., 2017; Kuuluvainen et al., 2021; Larocque et al., 2024), resulting in multiple overlapping disturbances over shorter-time periods, shifts in forest composition, and altered ecosystem stability (Jõgiste et al., 2017).

Clearcutting significantly alters forest macrohabitat structure by removing canopy cover and fragmenting continuous forest into smaller, disconnected, patches. This fragmentation increases the density of forest edges and open-canopy areas with limited shade, thereby reducing connectivity between remnant forest patches (Tuff et al., 2016). The resulting landscape is a mosaic of forest interior, edge, and open (cut) areas, each exhibiting distinct environmental characteristics (Bergeron & Fenton, 2012; Boucher et al., 2011; Remmel et al., 2023). Forest interiors generally exhibit greater heat storage capacity than open or edge environments due to

their dense vegetation, enclosed air space, and higher humidity levels, which result from reduced direct solar radiation (Geiger et al., 1965). In contrast, clearcuts and forest edges are subject to more extreme climatic fluctuations, often experiencing higher temperatures, lower humidity, and increased wind exposure (Chen et al., 1993; Murcia, 1995). Even-aged management practices, commonly associated with clearcutting, also influence long-term forest regeneration and structural complexity by simplifying age and species composition across regenerating stands (Bergeron et al., 1999; Fenton et al., 2009).

1.2 Modifications of microclimate from forestry practices

Macrohabitat modifications often lead to pronounced shifts in forest microclimates.

Microclimates refer to localized variations in physical conditions—such as temperature, humidity, light, and wind—that occur at small spatial scales within a larger habitat (Chen, 1999; Chen et al., 1993; Geiger et al., 1965). These fine-scale environmental conditions are shaped by structural features like vegetation density, canopy cover, and species composition, and they play a critical role in influencing species distributions and ecosystem dynamics under habitat change (Chen, 1999; De Frenne et al., 2019; Dobrowski, 2011; Keppel et al., 2017; Máliš et al., 2023).

Within forests, microhabitat variability occurs along both horizontal (edge-to-interior) and vertical (canopy-to-ground) gradients. Wind speed decreases with height due to surface friction; the ground absorbs solar radiation during the day and emits infrared radiation at night, contributing to daily microclimatic variation (Geiger et al., 1965). Canopy and ground vegetation play essential roles in moderating temperature and humidity through shading, evaporative cooling, and insulating ground heat at night (Burrow et al., 2023; Wolf et al., 2021). As one moves toward the forest floor, air temperature typically decreases while relative humidity

increases (Geiger et al., 1965). Together, canopy and ground vegetation help create more thermally stable forest microclimates (Burrow et al., 2023).

Microhabitat conditions may differ substantially between clear, partially cleared, edge, and forested plots (Chen et al., 1993; Geiger et al., 1965). Changes to ground level vegetation and woody debris (providing damp and shaded spaces), as well as mid-story foliage and canopy cover can modify the conditions of various microclimates by simultaneously altering incident solar radiation, albedo (surface reflectance), wind speed, decreased humidity, and resulting vapor pressure deficits (Campbell & Norman, 1998; Geiger et al., 1965; Lindenmayer et al., 2020; Murcia, 1995; Wolf et al., 2021). Canopy and understory layers regulate solar exposure by direct interception and through reflection from leaf surfaces, helping buffer extreme temperatures and reduce moisture loss through evaporative cooling (Wolf et al., 2021). Ground level vegetation further stabilizes temperatures by reducing evening temperature declines through insolation of ground radiation and retaining radiant heat (Geiger et al., 1965). Clearcuts are more spatially homogenous than unharvested areas due to the absence of layered vegetation and canopy structure (Lundmark et al., 2017; Rittenhouse et al., 2008). Yet, these open environments may experience greater diurnal variation, with sharper temperature drops at night and higher thermal extremes during the day compared to forested areas (Geiger et al., 1965).

1.3 Hydrothermal impacts on organismal performance

To successfully occupy a habitat and obtain energy needed for growth, maintenance, and reproduction, organisms must maintain a positive energy balance between themselves and their environment (Spotila et al., 1992; Tracy et al., 2010). Organisms exchange energy between themselves and their environments in the form of heat and water in multiple ways, including absorbed and emitted radiation, convection (through the surrounding air), conduction (through

surfaces), evaporation (Spotila et al., 1992). In ectotherms, where metabolic heat is minimal, the outcome of these exchanges determines body temperature and hydration status, both of which have high ecological relevance as individuals cannot maintain long-term activity in habitats where their body temperature surpasses critical limits or their daily water loss exceeds their daily water uptake (Spotila et al., 1992). Thus, the suitability of an environment for an organism is, in part, determined by the time in which it can remain active before reaching its maximum tolerable body temperature and dehydration level (Campbell & Norman, 1998). Altered microclimatic conditions can impact heat and water exchange processes and subsequent energy balances, thereby influencing the physiological suitability of the given environment for a particular organism.

Among ectotherms, amphibians may be particularly sensitive to microclimate modifications because they have permeable skin. The easy movement of water across their skin means that amphibians can substantially lower their body temperatures through evaporative cooling in environments where thermal conditions approach thermal limits, however this mechanism comes with a trade-off via increased vulnerability to desiccation (Spotila et al., 1992; Tracy, 1976). Vegetation loss, particularly of canopy and ground cover, exposes amphibians to increased solar radiation and higher vapor pressure deficits, which accelerate water loss and reduce soil and atmospheric moisture availability (Geiger et al., 1965; Wolf et al., 2021). Elevated wind speeds in disturbed forests can further dehydrate microhabitats by thinning boundary layers and displacing humid air, compounding stress on moisture-dependent species (Spotila et al., 1992).

Amphibian physiological performance is strongly shaped by body temperature and hydration state (Huey et al., 2012). Many key functions, including locomotion, respiration, and

growth, are sensitive to changes in temperature (Deutsch, 2008; Huey & Kingsolver, 1989), and dehydration can further constrain performance across these systems (Anderson & Andrade, 2017; Gatten, 1987; Greenberg & Palen, 2021; Preest & Pough, 1989; Rogowitz et al., 1999). Among these, locomotion provides an especially informative measure of performance because it integrates neuromuscular, metabolic, and behavioral responses to environmental stressors and is affected by organisms' hydrothermal state (Navas et al., 2008). Locomotion is necessary for processes affecting amphibian survival including reproductive efforts, predator avoidance, predation, and dispersal between habitats (Gatten et al., 1992). The relationship between performance and temperature is commonly represented as a thermal performance curve (TPC). TPCs generally show performance initially increasing slowly a maximum at the thermal optimum, then rapidly declining (Angilletta, 2009) (Figure 1.1.1). TPCs are not fixed and can be modified by an individual's acclimatization to previous or current environmental experiences (Schulte et al., 2011).

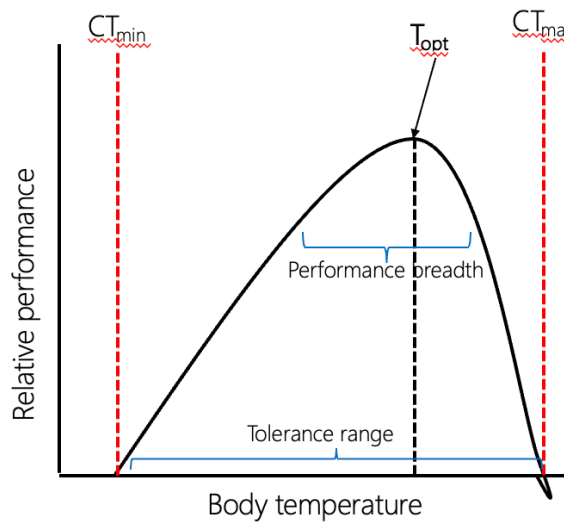


Figure 1.1.1 Hypothetical thermal performance curve including critical thermal minimum (CT_{min}), critical thermal maximum (CT_{max}), thermal optimum (T_{opt}), performance breadth, and tolerance range.

Dehydration can also alter thermal performance curves by shifting thermal optima downward (Anderson & Andrade, 2017; A. Mitchell & Bergmann, 2016; Preest & Pough, 1989), and narrowing the temperature range over which amphibians maintain high physiological function (i.e. performance breadth). While mild dehydration may have limited effects, performance declines sharply beyond this threshold, with frogs exhibiting reduced peak capacities and narrower thermal windows under moderate to severe dehydration (Anderson & Andrade, 2017; A. Mitchell & Bergmann, 2016). Therefore, thermal and hydric effects on performance are not independent and impacts of temperature on performance tend to be greater when animals are dehydrated (Greenberg & Palen, 2021). The combined effects of temperature and hydration state can be represented by a hydrothermal performance curve (HTPC). Multiple hydrothermal performance curves can be combined into a hydrothermal performance surface (HTPS).

Hydrothermal stress impairs amphibian performance by disrupting key physiological systems involved in locomotion, muscle function, and energy production (Anderson & Andrade, 2017; Greenberg & Palen, 2021; Moore & Gatten Jr, 1989; Rome et al., 1992). Amphibians can move in a variety of different ways for various tasks (e.g. reproduction, foraging, predator avoidance) through different types of muscle contractions, muscle fiber recruitment, and aerobic and anaerobic activity pathways (Rome et al., 1992). Aerobic capacity determines muscular endurance and controls activation and relaxation of muscles during sustained movement, while ATP and phosphocreatine stores are used for short-burst locomotion and maximal mechanical power output (Moore & Gatten Jr, 1989; Rome et al., 1992). Hydrothermal stress and desiccation impair aerobic function and deplete ATP stores, increasing reliance on less efficient anaerobic metabolism (Gatten et al., 1992; Preest & Pough, 1989). Cardiovascular performance declines with reduced heart rate, blood flow, and oxygen transport (Gatten, 1987; Gatten et al., 1992; Hillman, 1978). Dehydration-induced hyperosmolality draws water from muscles and the brain, accelerating fatigue (Hillman, 1978; Moore & Gatten Jr, 1989; Pough et al., 1983; Rogowitz et al., 1999; Senzano & Andrade, 2018). Reduced oxygen and hydration impair muscle fiber recruitment and timing of contractions (Rome et al., 1992), shifting metabolic activity towards less efficient anaerobic pathways. Although anaerobic pathways may sustain brief locomotion under stress, they yield less ATP and cause lactate buildup, accelerating time to fatigue (Gatten, 1987; Moore & Gatten Jr, 1989). In amphibians, intracellular ATP stores are sufficient to sustain maximal muscular effort for only 1–3 seconds, after which energy must be supplied by phosphocreatine and anaerobic glycolysis (Bennett, 1978). Consequently, anaerobic capacity can typically support intense activity for 30–60 seconds before fatigue sets in. Anaerobic capacity may be unaffected by small amounts of short-term dehydration and may allow individuals to

sustain brief periods of locomotion under hydrothermally stressful conditions. However, it remains unclear how ATP use and anaerobic locomotor capacities will be affected in amphibians experiencing frequent hydrothermally stressful conditions. The short-term limitations on anaerobic activity and the effects on aerobic metabolic pathways may be partially responsible for the reduction in locomotor capacities of dehydrated frogs (Gatten, 1987; Moore & Gatten Jr, 1989; Rogowitz et al., 1999).

1.4 Interspecific variations in hydrothermal tolerance

Individuals in different habitats experiencing changing hydrothermal conditions must balance their physiological tolerance margins with resource habitat preferences (Köhler et al., 2011). Because amphibians are often exposed to fluctuating microclimates, they routinely operate under varying hydration states and temperatures (Anderson & Andrade, 2017; Greenberg & Palen, 2021; Lertzman-Lepofsky et al., 2020; A. Mitchell & Bergmann, 2016; Navas et al., 2008). Environmental trade-offs can thus arise, where selecting optimal temperatures may increase desiccation risk and vice-a-versa. The nature and severity of these trade-offs vary across species and habitats, and the degree to which individuals can buffer hydrothermal microhabitat fluctuations depends partially on physiological traits like cutaneous resistance, body size, skin secretions, and oxygen uptake pathways (Gunderson & Stillman, 2015; Navas, 1996; Rogowitz et al., 1999; Spotila et al., 1992). Behavioural buffering strategies, such as seeking out refugia, adjusting posture, reducing activity, or altering movement patterns can also help minimize exposure to hydrothermal conditions near or above individual limits (Gunderson & Stillman, 2015; Spotila et al., 1992).

Species vary in their capacity to buffer hydrothermal stressors, and these differences are often shaped by both physiological tolerance limits and behavioral plasticity (Nowakowski et al.,

2018; Prates & Navas, 2009; Toledo & Jared, 1993; Wilson, 2001). High temperatures may be favourable and benefit performance for terrestrial and arboreal amphibians who possess physiological and behavioural adaptations that lower their desiccation risk (Hossack et al., 2009; Tracy et al., 2010). For instance, arboreal frogs with higher resistance to cutaneous water loss (Rc) than more aquatic species, aided by thicker skin and mucus secretions, can tolerate greater hydrothermal extremes and may engage more readily in behavioral thermoregulation (Roznik et al., 2018; Young et al., 2005). Arboreal species can also maintain favourable hydrothermal body conditions by moving vertically along trees and seeking out refuges located along trunks or branches to avoid increased temperatures at ground level (Biazzo & Quintana-Ascencio, 2022; J. R. Johnson et al., 2008). Terrestrial species may access standing water or moist refuges on the ground (e.g. under leaf litter along soil, in understory vegetation) if available (Hossack et al., 2013; Tracy et al., 2007). In contrast, species with lower dehydration tolerance are more restricted to favorable microhabitats and microclimates to maintain water balance (Navas et al., 2021).

The effectiveness and availability of behavioural modification strategies also depend on local habitat structure and quality and environmental context (Greenberg & Palen, 2021; A. Mitchell & Bergmann, 2016; Navas et al., 2008). The effects of altered temperatures or moisture availability are also context-dependent, such that changes in temperature and hydration may be positive or negative depending on whether they push organisms toward or away from their hydrothermal optima (Huey et al., 2012). High quality habitats (hydrothermally) allow organisms to operate close to their optimum for longer periods than low quality habitats with hydrothermal conditions near or above individual's tolerable limits. Activity in low quality habitats may be more energetically costly and impair an individual's ability to perform essential functions

required for survival, particularly if microhabitats offer limited buffering (Neel & McBrayer, 2018). Greater occurrence of low quality habitat requires animals to spend more time and energy behaviourally hydro-thermo-regulating by seeking refugia or maintaining water conservation poses during forced inactivity (Biazzo & Quintana-Ascencio, 2022; Navas et al., 2008; Sears et al., 2016; Sears & Angilletta, 2015), potentially impacting ecological functioning, fitness and even population dynamics. Improving our understanding of how environmental change and human activities affect hydrothermal physiological quality of habitats, and the shifting distribution of high- and low-quality habitat across the landscape is thus a key step in determining the vulnerability of species to ongoing habitat loss and modification.

1.5 Hypotheses

In this thesis, I determine how forest harvesting reconfigures the physiological quality of boreal landscapes for an arboreal amphibian. I examine microhabitat conditions between harvested and un-harvested forests at different stages of forest succession to determine if microhabitat conditions in clearcut environments reduce gray treefrog performance. These conditions are then compared with the species' hydrothermal niche determined from HTPS using jump performance. I propose the following hypotheses:

Hypothesis 1a: Initial impacts of harvesting on hydrothermal vulnerability are maintained during succession.

Rationale: Canopy removal in recent cuts is expected to increase exposure to temperature extremes both overnight and during the day and increases desiccation risk, limiting amphibian activity capacity. These differences may be maintained because of differences between clearcut regrowth and unharvested forest structure, since the former is shaped by intensive management

and develops high stand density with minimal canopy gaps, reduced sunlight penetration, and limited air movement, resulting in microclimates distinct from relatively unmanaged forests. These clearcut conditions that can persist for decades, with full recovery potentially taking a minimum of 60 years (Brassard & and Chen, 2006; Cyr et al., 2009).

- Predictions:

- 1) Performance will be lower in recent clearcuts than adjacent forest.
- 2) Performance will be significantly different between forest and clearcuts regardless of the time that has elapsed since cutting.

Alternative hypothesis 1b: Impacts of harvesting on hydrothermal vulnerability are rapidly erased by succession.

Rationale: This could be because rapid vegetation regrowth increases lower-level vegetation cover and shading, while reducing air movement due to stand density. These changes can enhance local humidity and create more favourable microhabitats for amphibians, leading to conditions in older clearcuts that are similar to those in adjacent forests

- Prediction: Differences in performance between clearcuts and adjacent forest will only be present in recent clearcuts.

Hypothesis 2: Clearcutting has a greater effect on hydrothermal vulnerability during nighttime activity than during daytime inactivity in disturbed environments.

Rationale: I expect this because during the day, frogs may make use of covered refuges to buffer against temperature variations and desiccation risk from exposure. However, overnight during peak activity hours, frogs may occur more frequently in uncovered microhabitats, which could expose them to rapidly cooling temperatures that approach their critical thermal minimums

- Prediction: Performance reductions between clearcuts and forests will be more pronounced overnight in exposed microhabitats than during the day in covered microhabitats.

2 METHODS

To understand how temperature and hydration jointly shape performance capacities in clearcut habitats, I first fit separate temperature-dependent and hydration-dependent performance curves, then combined them to generate a predictive hydrothermal performance surface across the full combination of laboratory-tested temperatures and hydration combinations. I then applied this predictive surface to infer potential performance of critical microhabitats in clearcut environments across forest recovery stages.

2.1 Hydrothermal Performance Tests

2.1.1 *Animal collection and housing*

We collected twelve adult male Eastern gray treefrogs (*Dryophytes versicolor*; Hylidae) on May 30 and 31, 2024, from three ponds. Four were collected on the evening of May 30 from a pond located at the northern end of Dog Lake Road, ON. Eight were collected on May 31 from two breeding ponds at the Kamview Nordic Centre, ON. I only used males because of the difficulty in finding enough females to achieve an adequate sample size. Animals were brought to Lakehead University's Biology Aquatics Facility (BAF) and housed in individual glass terraria and acclimated for one week at $20^{\circ}\text{C} \pm 3^{\circ}\text{C}$ and $>70\%$ humidity in a 12:12 light-dark cycle. Each tank was lined with damp paper towels and contained two PVC pipe perches, a refuge and plastic aquarium vegetation for enrichment, and a water dish. Paper towel substrate was regularly dampened with de-chlorinated water and was changed during daily cleaning or

when soiled. Water dishes were changed daily. Frogs were fed 6 crickets every 3 days, along with a dietary supplement of either multivitamin, calcium, or calcium +D3.

2.1.2 Performance testing schedule

I separated frogs into three groups of four individuals and randomly selected one frog from each group as a control throughout each of the trials. For all trials and groups, the control was kept at the experimental temperature but was not dehydrated and jumped on the same schedule as test animals to account for potential performance losses due to cumulative physiological fatigue over the course of each trial (Greenberg & Palen, 2021). Prior to testing, individuals were fasted for 24 hours to avoid digestion-induced thermal variation (Preest & Pough, 1989;). They then partook in one trial and were given a two-day recovery period before the next trial. One group was tested each day and the order of test temperatures was randomized across groups. At the start of each trial day, I randomized the order in which animals underwent trials. Start times were staggered every 30 minutes between animals to avoid having multiple individuals reach the target dehydration levels simultaneously. If, at any point during the trials, an animal was deemed unfit to continue, they were immediately removed from trials for the day and placed in a recovery water bath for at least an hour or until they had returned to their pre-trial weight and were returned to their enclosure. Veterinary approval was obtained before using these animals in subsequent trials (this occurred only once across all trials).

2.1.3 Hydration and temperature manipulations

Jumping ability was tested at four hydration levels (100%, 92%, 84 %, 75%, 70%) and five temperatures (37°C, 33°C, 24°C, 15°C, 8°C). Frogs were only taken to 37°C at 100% hydration. Body temperature and hydration status were varied by placing animals in a

dehydration tunnel (transparent acrylic tube with screened ends and a fan; design followed Greenberg & Palen, 2021) within an environmental chamber (VEVOR Reptile Incubator) set at the target testing temperature. Prior to serial dehydration, animals were placed within a water bath inside the environmental chamber set at the target temperature for 1 hour to reach full hydration. The environmental chamber remained at the target temperature for the duration of the trials. Initial frog weights were taken following the first set of jumps at full hydration. Dehydration levels were calculated as percent mass lost from their initial mass under the assumption that mass lost would be almost entirely water. Individuals were then placed within their respective tunnels and weighed every five minutes to monitor weight loss until the target dehydration was reached. Serial dehydration, rather than a random, approach reduced the number of times animals were dehydrated, and was a necessary animal welfare consideration (Greenberg & Palen, 2021). There were three occurrences during which frogs were not taken to 70% hydration. The first was on the second day of trials at 15°C when frog A2 had been dehydrating for over 5 hours and was only at 84% hydration. We stopped trials at 75% as we were worried about overly stressing the frog given the long dehydration period. Subsequently, if frogs dehydrated slowly (< 0.01 g every five minutes for over 30 minutes) we added an extra fan to the dehydration tunnel. The second occurrence was when A1 obtained a minor injury and was removed from trials for the day; he was subsequently examined and cleared for a return to trials by a veterinarian. The third occurrence was when frog A6 did not jump during his trial of 24°C at 75% hydration despite provocation and so was not dehydrated to 70% for that temperature.

2.1.4 Jump testing

Once animals reached the target dehydration, they were removed from their tubes with damp gloves and placed into the jumping arena. The arena was a 122cm x 200cm white MDF board

surrounded by an 80 cm high plastic wall (Figure 2.1). A GoPro Hero 11 mini (linear, 4K, 60 fps) was fixed 120 cm above the arena floor. Scale bars 100 cm and 97 cm in length were affixed to the arena floor in the x and y directions. If animals did not jump immediately, they were encouraged to by feigning capture by hand from behind or lightly prodding their urostyle. After a maximum of 6 jumps, they were placed back into their containers and dehydration continued until they had reached the next dehydration target.

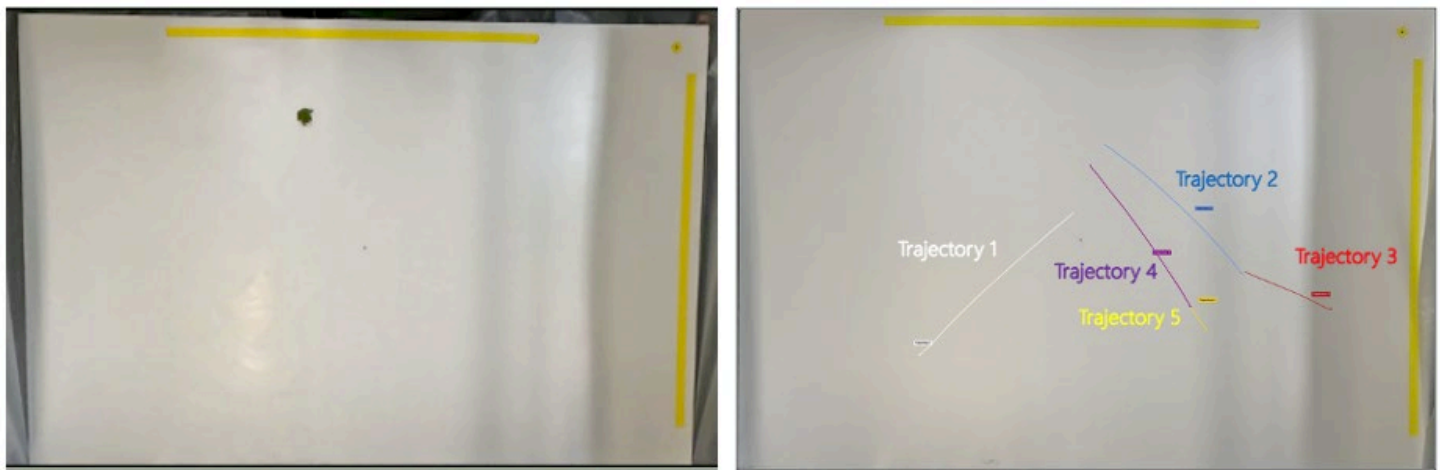


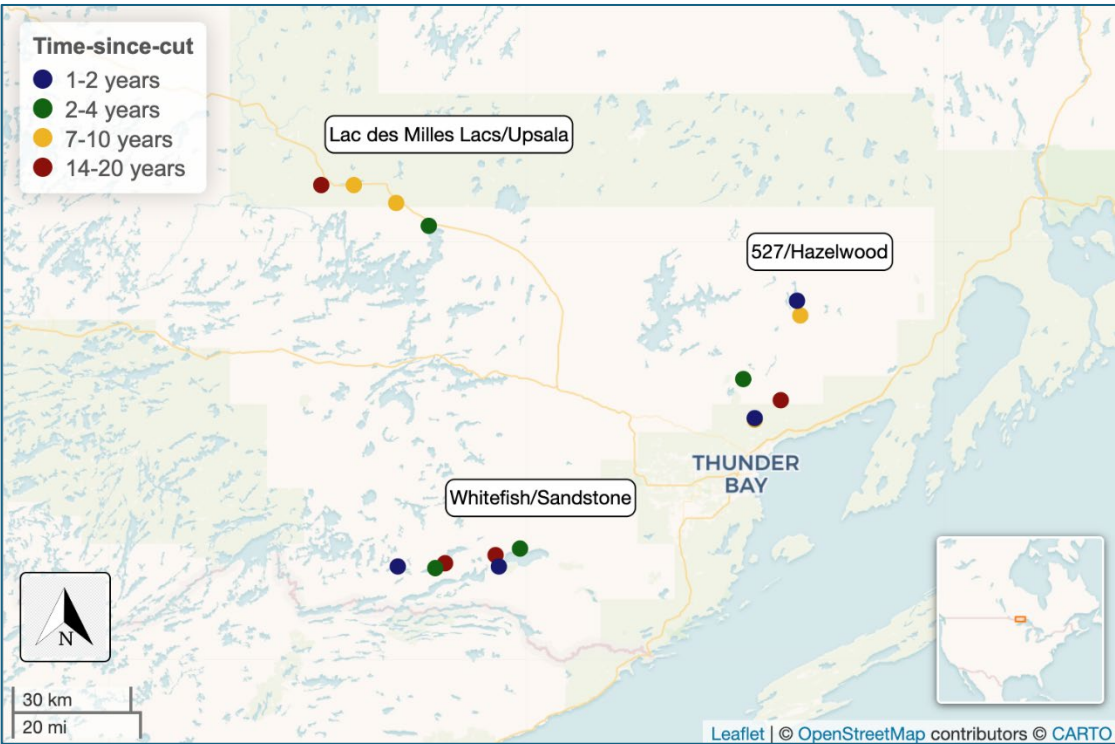
Figure 2.1. Jump arena set up. Platform was a 122cm x 200 cm white MDF board surrounded by 80 cm high plastic wall. A GoPro Hero 11 mini (linear, 4K, 60 fps) was fixed 120 cm above the arena floor. Pieces of tape 100 cm and 97 cm in length were affixed to the arena floor in the x and y directions for scale bars respectively. Left image is clear board with a frog placed away from the center. Right image is a screenshot of one trial with labeled trajectories.

2.2 Field Microhabitat Data collection

2.2.1 Study Area and Site Selection

To study the effect of clearcutting and succession on microhabitat quality for gray treefrogs, I used 16 sites split across three areas in the boreal forest transition zone around Thunder Bay, Ontario, Canada: 134 km northwest of the city, near Lac des Milles Lacs/Upsala;

394 40 km north of the city along Highway 527 and Hazelwood Conservation Area; and up to 120
395 km southwest of the city between Whitefish, Sandstone, and Sunbeam Lakes (Figure 2.2). All
396 field work was conducted in Northwestern Ontario in the boreal forest transition zone from July
397 to September 2024, during a portion of the active months for gray treefrogs in this region. All
398 sites were within the Pigeon River and Lake Nipigon Ecoregions.



399
400 **Figure 2.2** All sites used during the summer 2024 field season. There are three main locations
401 with respect to Thunder Bay: Lac des Mille Lacs (northwest) , Whitefish/Sandstone
402 (southwest), and 527/Hazelwood (north of Thunder Bay). Marker colours identify years of
403 succession, i.e. since cutting: 1-2 years (yellow), 2-4 years (green), 7-10 years (red), and 14-18
404 years (blue).

1-2 years



2-4 years



7-10 years



14-20 years



Figure 2.3 Vegetational regrowth across each time-since-cut in all cut sites. All photos were taken from eye height (168cm).

The northwestern Ontario section of the Boreal Shield transition zone consists of mixed-wood and boreal forests, with tree species including black spruce (*Picea mariana*), jack pine (*Pinus banksiana*), white spruce (*Picea glauca*), balsam fir (*Abies balsamea*), trembling aspen (*Populus tremuloides*), white birch (*Betula papyrifera*), and balsam poplar (*Populus balsamifera*) (Brandt et al., 2013). The region is further characterized by numerous lakes, and widespread wetlands such as bogs and fens (Crins et al., 2024). The climate is continental subarctic to humid continental, with long, cold winters and short warm summers. Summer months experience longer

day lengths due to higher latitudes of the boreal zone, experiencing on average 10-12 hours of sunlight (Price, 2013).

I identified all field sites using a combination of LANDSAT data, the Boreal Disturbance Database, and MNR forest management planning (FMP) maps (Hermosilla et al., 2016; Remmel et al., 2023; Wulder et al., 2024). Sites were controlled for relative size of disturbance and pre-disturbance forest type. Areas that had experienced multiple harvests throughout different years were not used due to lack of available uncut forest for controls and to avoid potential compounding effects from multiple continuous harvests (Anoszko et al., 2022). We restricted our study to seeded clearcuts rather than planted clearcuts because in Ontario planted sites are almost exclusively conifer-dominated, whereas seeded sites regenerate into deciduous stands. Including both would have introduced different regeneration pathways and structural trajectories and increased variation in our clearcut sites. All selected clearcut sites had undergone the same post-harvest regeneration treatments according to the OMNR forest inventory plans. However, data on tending, debris removal, and seeding was difficult to obtain for older sites (12+ years). I aimed to minimize differences within sites by identifying areas with similar terrain, slope, and forest type using LANDSAT data and site visits as per recommendations in deMaynadier & Hunter (1995). All cut sites were rapidly colonized by dense stands of aspen (*Populus tremuloides*) and birch (*Betula papyrifera*) all within the same age (Figure 2.3) in contrast to the more compositionally diverse mixed-wood forests that include coniferous species such as balsam fir (*Abies balsamea*) and jack pine (*Pinus banksiana*), alongside native deciduous species.

436 2.2.2 *Study Design*

437 I implemented a crossed hierarchical design in which sites were nested within time-since-
438 cut group and plots were nested within sites. I considered four time-since-cut groups: 1-2 years,
439 2-4 years, 7-10 years, and 14-20 years. I used 16 sites in total (four time-since-cut group x four
440 replicates); each site was divided into three plots: control (forest not cut recently), edge, and
441 harvested. In each plot, I estimated adult treefrog operative temperatures and evaporative water
442 loss (EWL) using plaster models (Hastings et al., 2023; Peterman et al., 2013; Tracy et al., 2007)
443 in three microhabitats: ground, sheltered trunk, and open trunk (Dodd, 2013; J. R. Johnson et al.,
444 2008). Each site was sampled twice over the season with a minimum of 3 weeks between sample
445 periods (Table 2.1).

Table 2.1 Sites used for sampling with geographic location, sampling dates, region, and time-since-cut groupings. Date ranges indicate that sampling was done over two days for that period to encompass both daytime and evening collection periods.

Site Name	Region	Time-since-cut	Longitude	Latitude	Sampling Dates
A2008	527/Hazelwood	14-20 years	-89.145	48.572	Jul 8-9 and Aug 26, 2024
A2015	527/Hazelwood	7-10 years	-89.075	48.760	Jul 4 and Aug 6-7, 2024
A2022	527/Hazelwood	1-2 years	-89.087	48.791	Jul 3 and Aug 27, 2024
H2016	527/Hazelwood	7-10 years	-89.230	48.526	Jul 27 and Aug 30, 2024
H2020	527/Hazelwood	2-4 years	-89.268	48.618	Jul 5-6 and Aug 28-30, 2024
H2022	527/Hazelwood	1-2 years	-89.230	48.532	Jul 31- Aug 1 and Sep 1, 2024
L2010	Lac des milles Lacs/Upsala	14-20 years	-90.694	49.049	Jul 18-19 and Aug 13, 2024
L2014	Lac des milles Lacs/Upsala	7-10 years	-90.438	49.007	Jul 30-31 and Aug 11, 2024
L2015	Lac des milles Lacs/Upsala	7-10 years	-90.585	49.049	Jul 19-20 and Aug 12, 2024
L2021	Lac des milles Lacs/Upsala	2-4 years	-90.330	48.957	Jul 17 and Aug 10, 2024
W2006	Whitefish/Sandstone	14-20 years	-90.276	48.208	Jul 2-3 and Aug 21, 2024
W2009	Whitefish/Sandstone	14-20 years	-90.104	48.223	Jul 24-25 and Aug 20, 2024
W2020	Whitefish/Sandstone	2-4 years	-90.021	48.241	Jul 23-24 and Aug 15, 2024
W2020B	Whitefish/Sandstone	2-4 years	-90.308	48.195	Aug 4-5 and Aug 22, 2024
W2022	Whitefish/Sandstone	1-2 years	-90.435	48.197	Jul 22-23 and Aug 17, 2024
W2023	Whitefish/Sandstone	1-2 years	-90.091	48.200	Jul 3-4 and Aug 23-24, 2024

2.2.3 Site Preparation

Trees were randomly selected within each cut-block section of the site and then tree species, tree height and diameter at breast height (DBH), and distance between trees was replicated, as closely as possible, in the edge and forested plots. Tree selection was limited by tree availability in harvest plots, as there were generally fewer large standing trees in those environments, particularly in older plots. However, I aimed for two coniferous and two

deciduous in each plot. Edge habitat was considered to be within 30 meters on both sides of the treeline (Boucher et al., 2011; Matlack, 1993). I recorded site-characteristics around each tree including ground, mid-story, and upper story vegetation, canopy cover above replica placement, and leaf litter quantity. I measured heights for each tree whose microhabitats were being measured. I recorded overall macrohabitat weather (temperature, relative humidity, and dewpoint) for each clearcut with a temperature logger (HOBO Onset MX2301A) paired with a radiation shield for each recording period.

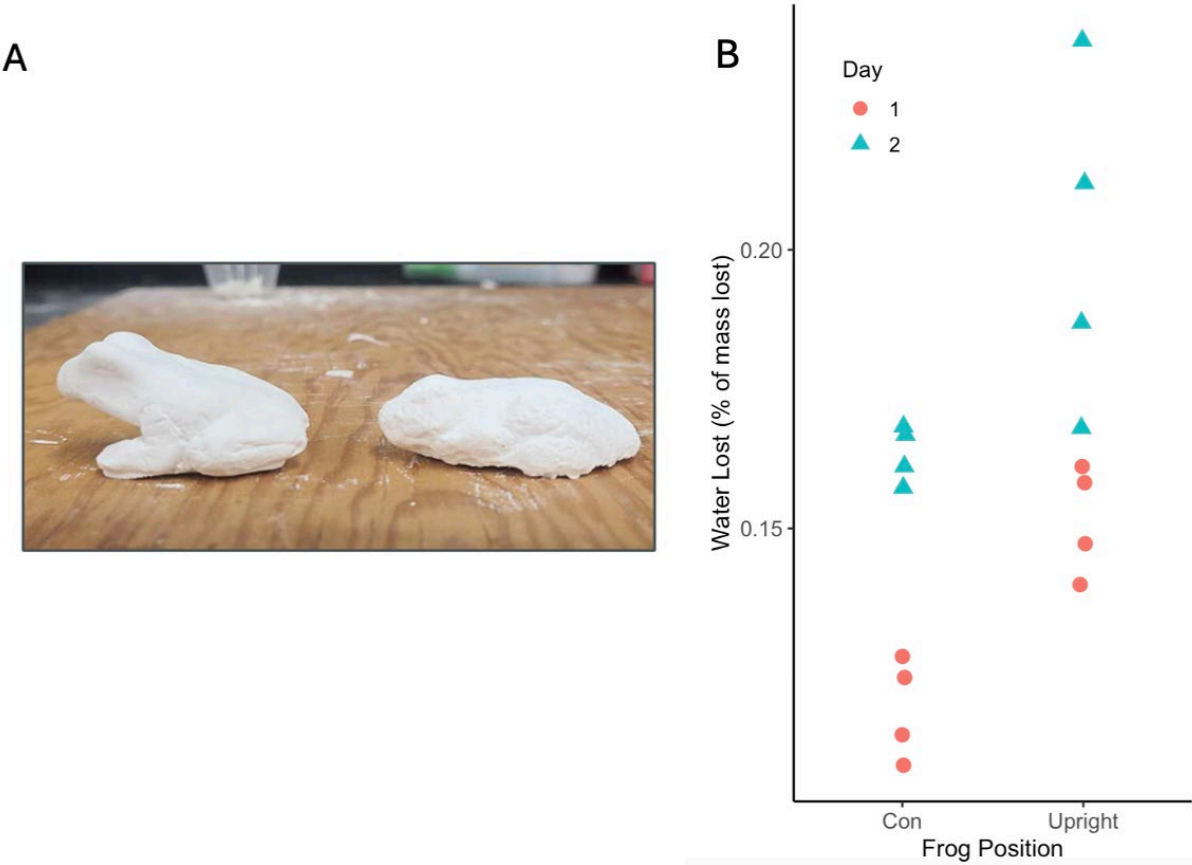


Figure 2.4 Sampling of macro- and micro-habitats at a single site. Left: an idealized site with three macrohabitat plots: Forest, Edge, and Cut. Middle: replica deployment on each tree, with (1) uncovered replicas, (2) replicas covered with pvc pipe, and (3) ground replicas. Right: photos of each microhabitat at a 2022 cut site.

2.2.4 Replica construction and calibration

I used plaster replicas (Hastings et al., 2023; Peterman et al., 2013; Tracy et al., 2007) to measure evaporative water loss (EWL) and surface temperatures in three microhabitats: trunk-covered, trunk-uncovered, and ground. Replicas were based on Tracy et al. (2007). Plaster

473 replicas were created in two postures that real frogs adopt: water conservation (inactive) and
474 upright (active) (Figure 2.5). Prior to field deployment, I tested whether plaster replicas in
475 upright postures lost more water than those in water conservation. I used five replicas for each
476 posture and tested them over two days. Each of the models were weighed, soaked in water for an
477 hour, re-weighed and then left for four hours at 21°C in front of a fan to simulate the time they
478 will be left in the field. They were then re-weighed to determine water loss as a percentage of
479 original body mass. I found upright models to lose water faster than flat models and thus used
480 both postures for all subsequent work. In total, I built 300 replicas, 150 in each posture.



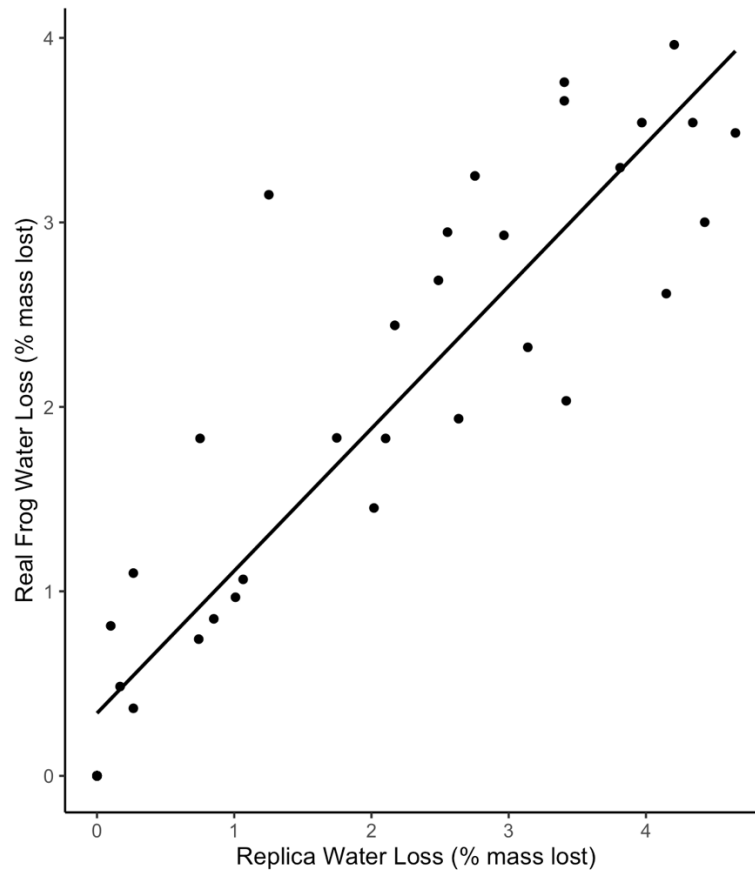
481

Figure 2.5 Evaporative water loss from plaster replicas in two postures at 22.5 °C. A: example replicas in upright (left) and water conservation/flat (right) postures. B: Differences in water loss between postures (n=4 replicas in each posture, with water loss measured across 2 days). The plot indicates water loss measurements taken over two days.

To determine if my replicas approximated EWL rates of real frogs, I conducted calibration tests using three of the live frogs captured for the jumping performance tests, three upright replicas, and three flat replicas. These tests were conducted after jump trials had concluded and experimental animals had rested for 2 days. Each frog (real and plaster) was soaked in water for 60 minutes to reach full hydration. They were removed from their water, weighed, brought outside and placed in mesh-covered dishes in direct sunlight. Every five minutes I recorded surface temperatures (using Omega Infrared thermometer), weight, and whether the real frogs were in water conservation. Mass lost every five minutes was calculated by subtracting each weight at five minutes from the initial full hydration weight to get a total weight lost at the end of the 70-minute period. I compared the rate of weight loss between real frogs in upright position with replicas in upright position, and between real frogs in water conservation with replicas in water conservation position (Figure 2.6). I found a significant positive correlation between rates of real frogs and replicas ($r = 0.865$, $p < 0.0001$). I did not perform temperature corrections to match dehydration corrections due to lack of data on internal body temperatures of frogs during correction testing. To calculate the correction between real and replica frog water loss rates, used the following formula:

$$CF = \frac{\beta_1}{\beta_2}$$

503 Where: CF is the correction factor, β_1 is the slope of linear regression model for treefrog water
 504 loss against time; and β_2 is the slope of the linear regression model for replica water loss against
 505 time.



506
 507 **Figure 2.6** Water loss rates of real frogs (n=3) and plaster replicas (n=3) measured every five
 508 minutes over 70 minutes in an identical outdoor environment exposed to natural weather
 509 conditions (e.g. wind, radiation, humidity, temperature). Black line indicates the linear model for
 510 real frog water loss ~ replica water loss.

511 2.2.5 Field replica deployment

512
 513 All replicas on tree trunks were placed along the north-facing side of trunks 2 metres
 514 from the ground, and ground replicas were placed directly below also on the north-facing side
 515 around crevices or under vegetation on the ground. Trunk shelters were made from grey PVC

pipe to standardize shelter size and structure. Two replicas of different postures were paired together for each microclimate measurement (Figure 2.5) to estimate conditions at different activity levels in the field. Each pair of replicas was placed as closely together as possible (within 5-6 cm) with a HOBO MX2201 or HOBO pendant temperature datalogger in between them. Out of 16 sites, there were four from the early successional stages at which replicates in the harvest plots were placed on two adjacent trees of the same trunk size and height as single trees were too small to accommodate both models. All other sites had replicas on the same tree. Trunk replica pairs were aligned vertically and ground replica pairs side-by-side (Figure 2.4; right). Vertical positioning of replicas (Figure 2.4) was randomized at each site to account for effects of placement on water loss were.

For field deployment, replicas were soaked in water for one hour (time to full saturation), then weighed prior to deployment. I placed replicas at the three microhabitats within each plot (one site x three macrohabitats x three microhabitats x four replicates of each microhabitat = 48 replicas at each site). Replicas were left for four hours during the day or 11 hours at night in each plot before being re-weighed to estimate evaporative water loss. Night and day deployment times differed because I assumed that replicas left out for 11 hours during the day would desiccate completely and I would be unable to determine when 100% water loss occurred. I recorded temperature experienced in microhabitats every 30 minutes as closely as possible to each pair of frogs using an unshielded temperature logger (HOBO MX2201 or HOBO pendant) for both day and evening temperatures. Replica surface temperature during the day was recorded manually every 30 minutes using an Omega thermocouple RDXL4SD type T. Data was collected during the day from 10:00 to 14:00, and overnight from 20:00 to 7:30 the next day. Deployment times shifted by 30 minutes near the middle and end of August to match the shift in sunset and sunrise

times. To avoid bias and to minimize differences in deployment and removal times, the order of deployment within each plot was randomized and then maintained for removal to ensure replicas were deployed for equal amounts of time. Since plaster replicas eventually began to deteriorate, they were replaced when their dry mass was reduced to 80% of their original dry mass (Tracy et al., 2007).

2.3 Analysis

All statistical tests were performed using software R version 4.3.1 (R Development Core Team). I used packages nlme (Pinheiro et al., 2025) and glmmTMB (Brooks et al., 2017) for model fitting, DHARMA (Hartig et al., 2024), emmeans (Lenth et al., 2025), performance (Lüdtke et al., 2021), and afex (Singmann et al., 2024) for model checks and post-hoc analysis. I used the interp (Gebhardt et al., 2024) package for bilinear interpolation of performance estimates.

2.3.1 Jump Distance

All videos were analyzed using Kinovea™. I measured the distance from each jump's start to endpoint for each frog and trial combination and used only the maximum jump distance as an estimate of maximal performance (H. John-Alder et al., 1988). If a frog did not jump or move, jump distance was recorded as zero. Jumps were considered distinct from hops as a form of movement and were distinguished by full extension of the frogs' hind limbs (see Mitchell & Bergmann, 2016). Though acceleration is occasionally used as a measure of performance alongside jump distance, (A. Mitchell & Bergmann, 2016; Wilson, 2001) I did not calculate acceleration. Each maximum jump distance was used to fit hydro- and thermal- performance curves in section 2.3.2.

I used a linear mixed effects model to test the effect of the time frogs spent in a trial on jump distance, with fixed effects of time under trial and random effect of frog and date. I conducted this test to isolate potential physiological fatigue from being in trials from the effects of dehydration by comparing control frogs that did not undergo dehydration with test frogs that did undergo dehydration. I found no significant effect of time under trial on control frog jump distance ($\beta = -0.00479$, $df = 48$, $SE = 0.0078$, $p = 0.541$; Figure 2.7).

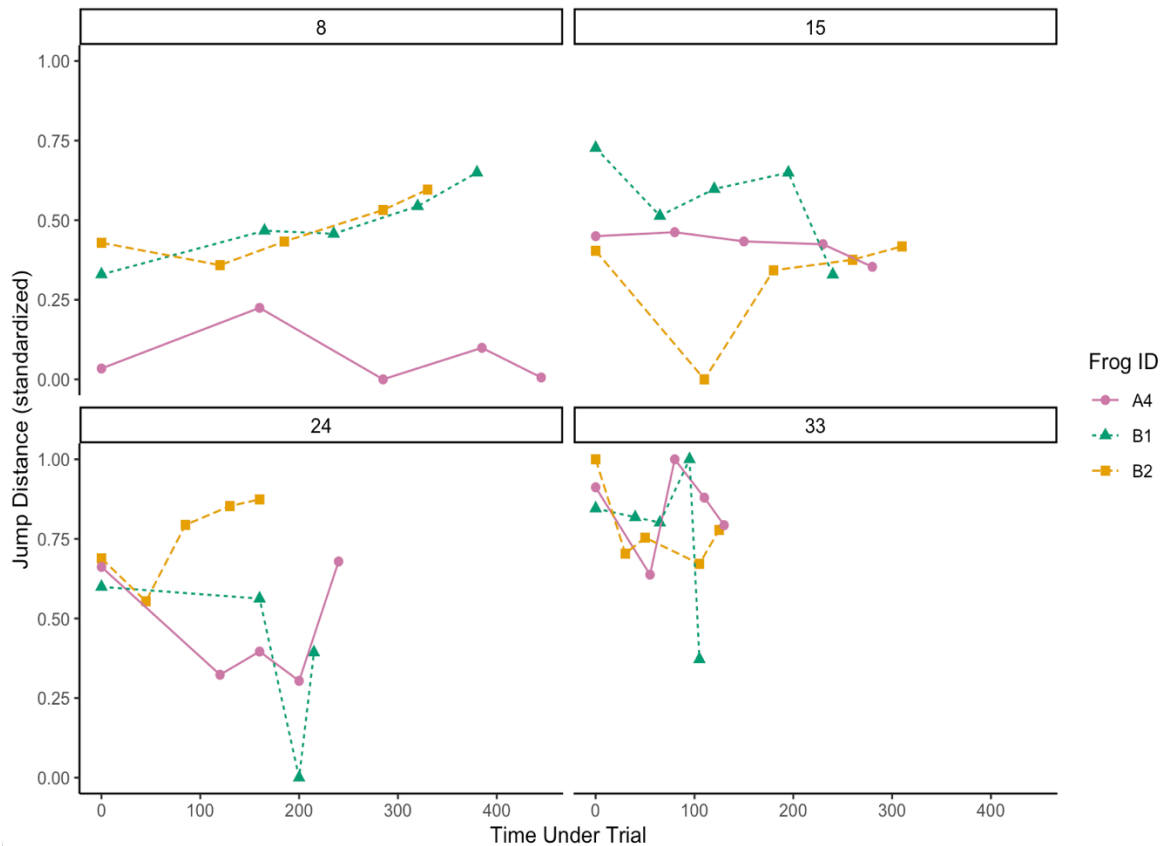


Figure 2.7 The effect of time under trial on jump distances for control frogs. Each plot is grouped by trial temperature. Points indicated normalized jump distances for each of the control frogs.

2.3.2 Hydrothermal Performance Curves

To quantify hydrothermal performance relationships for gray treefrogs, I fit a candidate set of thermal performance functions and dehydration-performance curves to laboratory jumping

data (full list in Appendix B) for each frog separately. To account for uncertainty in model selection, I ranked all fitted candidate models for each frog using Akaike's Information Criterion (AIC), calculated Akaike weights, and generated model-averaged predictions for that individual. This produced one averaged curve for each frog for each axis (temperature or hydration). I modeled thermal and hydric performance relationships separately (univariate fits) because (1) no established functional form exists for a joint temperature–hydration performance curve (Angilletta, 2009; Huey & Stevenson, 1979), and (2) my dataset did not support fitting high-dimensional nonlinear mixed models without convergence failures or biologically unrealistic fits. Fitting models ignoring that each individual was measured multiple times was not an option as it would result in pseudoreplication.

Fitting curves separately for each frog avoided problems with mixed-effects nonlinear model structure and allowed me to retain the shape and scaling of each individual's response (Figure 3.1 and Figure 3.3). Within each frog, I fit candidate functions (adapted Kontopoulos et al., 2024 and Padfield et al., 2021) with nonlinear least squares regression (Appendix B) with either hydration or temperature as a response. I scaled jump distances to 0–1 (where 1 = that frog's maximum distance). I excluded fits if they were multimodal (> three peaks within the observed range), exhibited unrealistic monotonic change across the experimental range, or clearly overfit the data (Angilletta, 2006). I only used frogs that had at least three data points (maximum jump) for each temperature and dehydration combination. This meant that for 70% dehydration, there is no weighted curve for frog A6.

By using a broad candidate set for each univariate relationship and applying model selection and model-averaging, I aimed to reduce selection uncertainty in curve shapes (and therefore parameter estimates) and to ensure that predictions were not unduly influenced by a

single arbitrarily chosen model (mixed models selection: Angilletta, 2006, Dormann et al., 2018; J. B. Johnson & Omland, 2004; Kontopoulos et al., 2024; Rozen-Rechels et al., 2019).

Restricting the analysis to a single or small set of functions risks model misspecification, which could bias parameter estimates and predictions (J. B. Johnson & Omland, 2004).

For species-level predictions, I combined curves in two stages:

(1) **Within-frog averaging:** Model-averaged curves were generated using Akaike weights, so the contribution of each candidate function reflected its relative support for that frog's data (Appendix A for model weights).

(2) **Across-frog averaging:** The resulting individual model-averaged curves were then combined into a single population curve by equal-weighting each frog's contribution. This ensured that all individuals contributed equally to the species-level estimate, regardless of the number of candidate functions retained for each frog

I used the population model-averaged curves to build a predictive surface of jumping performance that combines temperature and hydration (Figure 3.5). To estimate field performance, I merged the two univariate surfaces into a bivariate surface by pairing predictions from the corresponding hydration and temperature curves and interpolating across the microhabitat field data (temperatures and hydration states) (**Figure 3.5**). The resulting predicted jump distances served as the response variable in subsequent statistical models comparing potential performance across sites and microhabitats (Section 2.3.3). This approach is analogous to using derived variables, common in species distribution or ecological niche modelling techniques, where predictions are treated as response variables in subsequent analyses (Elith & Leathwick, 2009).

I acknowledge several limitations in the above approach to curve fitting and interpolation. First, treating the curves univariately and later combining them may inflate the effective sample size, since the same individuals contribute to both fits. Second, propagating predictions from two separate models into a single combined surface inevitably carries forward the uncertainties from each fit. Nonetheless, given the current dataset and the study's primary objective of estimating performance consequences under realistic field conditions, this approach represents a compromise between model complexity and data limitations. While this method allows for an integrated link between measured field conditions and laboratory-based performance relationships, it should be interpreted cautiously, as the predictions are subject to the combined errors and assumptions of both datasets.

2.3.3 Hypothesis testing

To examine differences in site conditions between clearcuts, forests, and edges through forest regrowth, I used model-averaged predictions from Section 2.3.2 to estimate expected performance under field conditions, using observed environmental temperature and hydration data. In this section, I tested for differences between macrohabitats (cut, edge, forest), microhabitats (ground, covered, uncovered) across each time-since-cut age. For each hypothesis, I fit mixed effects models with either temperature, water loss, or performance as the response variable. Models were fit for day and night separately. All models included random effects of sites nested within date to account for site variation, weather differences among days, and temporal autocorrelation. I performed all analyses using replica temperatures averaged across the entire period of deployment, corrected water loss levels calculated using the correction factor derived from calibration tests (correction factor ~ 0.7400), and performance estimates derives from jump testing data described above (Section 2.3.2). Sites with a minimum of one rainy day

were removed from the water loss and performance models to allow for proper specification of the nesting structure for date of visit and corresponding sites. For models that demonstrated autocorrelation after including random date effect, I checked for stationarity using Augmented Dickey-Fuller (ADF) tests and used autocorrelation function (ACF) and partial autocorrelation function (PACF) plots to determine p and q values for correlation structures. I checked all final models for assumptions of residual overdispersion, heteroscedasticity, normality, and independence. I performed multiple comparisons of estimated marginal means (with Tukey corrected p-values for multiple comparisons) post hoc tests to analyze interactions and fixed effects of significant in each of the models.

2.3.4 Hypothesis one: macrohabitats through succession

For hypothesis one models with temperature as a response, I fit non-linear mixed effects models (nlme package) with gaussian distributions and autocorrelation specifications ($p=1$, $q=0$). For the responses of water loss and performance, I fit generalized linear mixed effects models with a t-family distribution. I fit all models testing hypothesis one with a fixed interaction between macrohabitat (cut, edge, and forest) and age, and main effects of macrohabitat, age, and microhabitat. The random effect was date nested within site (Figure 2.8).

Response: Water loss *or* Temperature *or* Predicted performance

Fixed Effects: Macrohabitat x Age + Microhabitat

Random Effects: (1 | site/date)

Figure 2.8 Hypothesis one model structure. Models were fit for each response variable and each period (day or overnight) separately. Models were fit as non-linear mixed effects models with a gaussian distribution and autocorrelation function specifications.

2.3.5 Hypothesis two: microhabitat and posture differences

For all models testing hypothesis two, I fit generalized linear mixed effects models with a t-student distribution (Brooks et al., 2017; glmmTMB package). I included the same fixed effects as for testing hypothesis one but added posture and interaction terms between microhabitat and age, and microhabitat and macrohabitat (Figure 2.9). Effects of posture on temperature could not be determined for overnight data as one temperature measurement was taken for each pair of upright and flat replicas in each microhabitat. I could not directly test the difference between day and night performance effects in the same model due to differences in time intervals for water loss rates. The same procedures and model checks were followed for all models as described above (section 2.3.3).

Response: Water loss *or* Temperature *or* Predicted performance

Fixed Effects: Macrohabitat : Age + Age : Microhabitat + Macrohabitat : Microhabitat + Macrohabitat + Age + Microhabitat + Posture

Random Effects: (1 | site/date)

Figure 2.9 Hypothesis Two Model Structure. Models were fit for each response variable and each period (day or overnight) separately. Models were fit as generalized linear mixed effects models when residual distribution required alternative specification (t-family)

In the above section, my aim was to test for differences in expected performance among habitats and conditions. I therefore used a frequentist approach (linear/mixed-effects models) for hypothesis testing. This is distinct from the statistical method used in Section 2.3.2, wherein I used information-theory to estimate hydrothermal performance curves with no *a priori* hypotheses about general curve shape. These two stages address different questions—curve estimation versus hypothesis testing—and thus employ the statistical framework most appropriate to each objective. Although the second stage is based on predictions from the first, I

treat them as conceptually distinct analyses. Using different statistical paradigms in different parts of an analysis is common in ecological modeling when stages have distinct inferential goals (Burnham & Anderson, 2004; J. B. Johnson & Omland, 2004). In such cases, the choice of framework is guided by the nature of the question being addressed, rather than by strict adherence to a single philosophical approach.

3 RESULTS

3.1 Thermal Performance Curves

Thermal performance curves revealed that both thermal breadth and thermal optima shifted with hydration level (Figure 3.2), with T_{opt} at 100% hydration substantially higher than at other hydration levels. Frogs reached their highest maximum performance at 92% hydration ($\max = 0.81$), and their lowest maximum at 70% hydration ($\max = 0.26$). Thermal performance breadth (defined as the range of temperatures where performance was 90% of maximum) narrowed only slightly; at 100% hydration, it spanned 11 °C while at 70% hydration it spanned 9 °C. However, its position varied greatly, being situated between 28°C and 39.0 °C at 100% hydration and from 22°C – 31 °C at 70%. Individual curves varied in shape and magnitude but consistently showed declines in performance at both high and low thermal extremes when dehydration increased. Individual weighted curves are present in Figure 3.1 demonstrate variation in thermal performance across frogs under different hydration treatments.

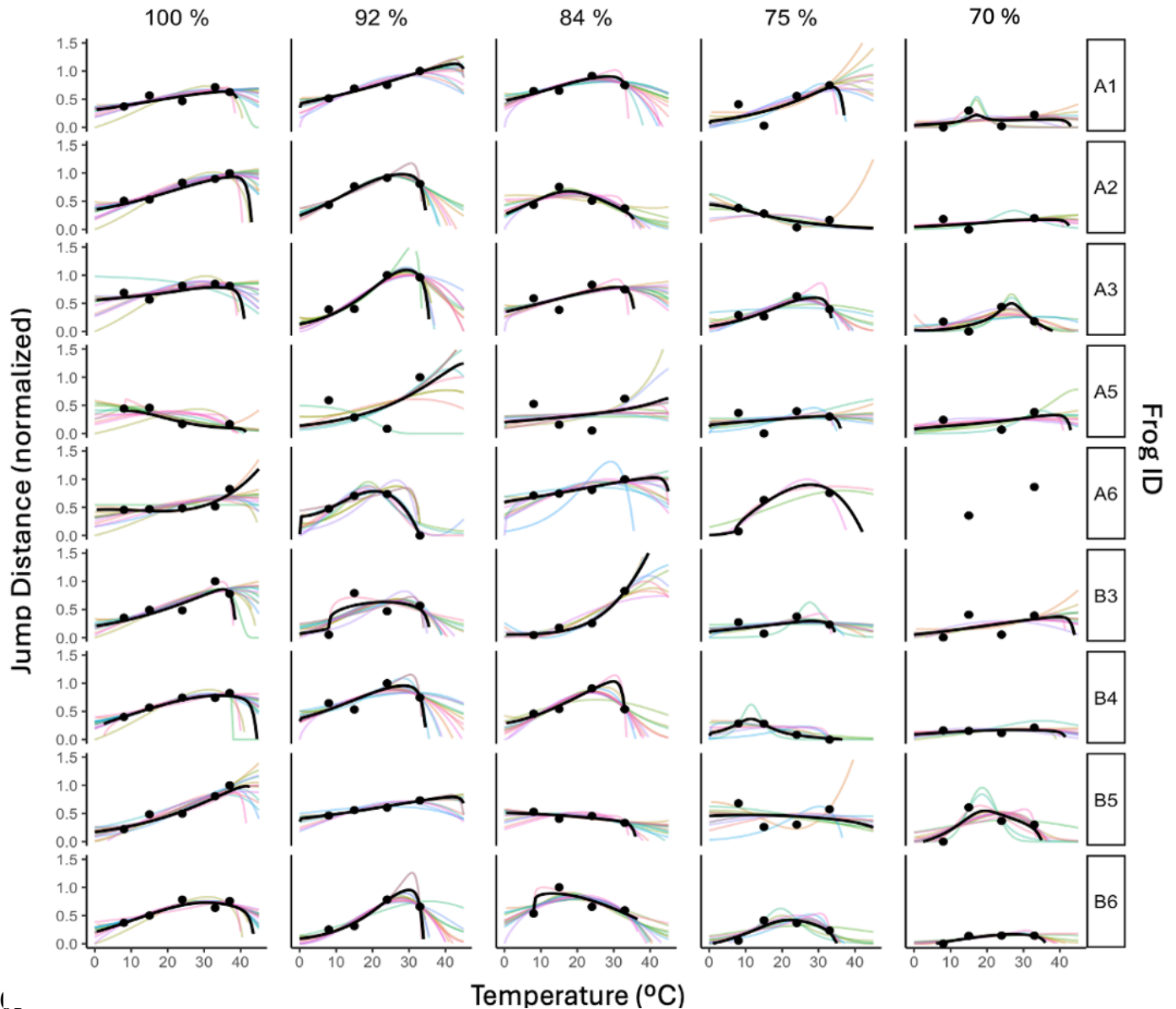
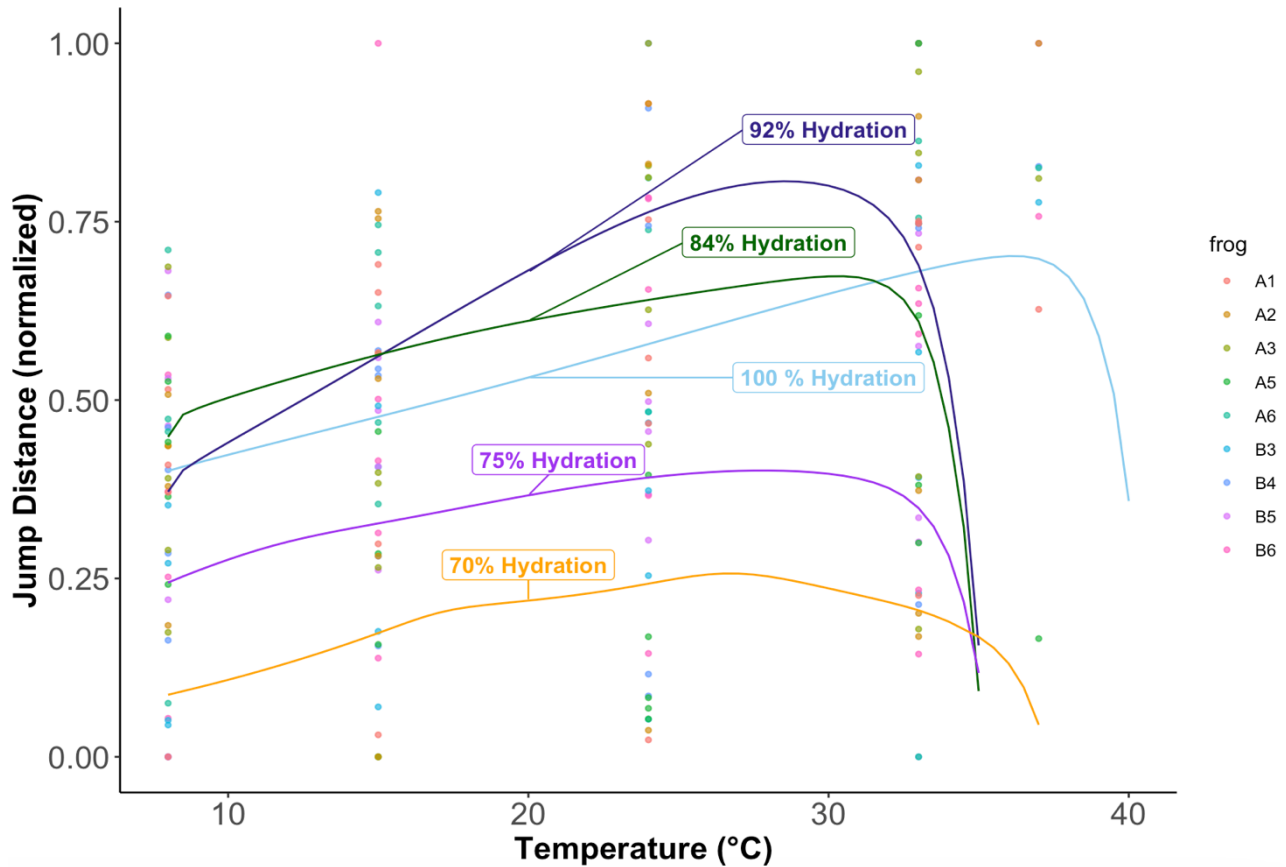


Figure 3.1 Weighted-average thermal performance curves for each frog at each hydration level. Faded coloured lines for each frog (each rectangle) indicate single curves fit from model types. Performance was measured as maximum jump distance, standardized as the proportion of each frog's maximum jump distance across all temperatures. Standardized performance values for each frog are indicated by black points. Temperature was measured from incubation temperatures in which frogs were held during trials. Solid black lines for each plot are results from Akaike weighted average curves for each frog. Not all models were fit for each frog due to parameter (k) restrictions and final curve shapes for each model.

714



715

716 **Figure 3.2** Thermal performance curves at five dehydration levels (100%, 92%, 84 %, 75%,
 717 70%). Data for trials was subset for dehydration levels across five temperatures (37°C, 33°C,
 718 24°C, 15°C, 8°C). Frogs were only taken to 37°C at 100% hydration. Performance was
 719 measured as jump distance, normalized as the proportion of each frog's maximum jump distance
 720 across all temperatures. Temperature was measured from incubation temperatures in which frogs
 721 were held during trials. Curves produced are weighted averages from the curves fit to each frog.
 722 Coloured points correspond to standardized maximum jump distance for frogs (A1, A2, A3, etc.)

723 3.2 Hydro-Performance Curves

724 Hydro-performance curves indicated peak performance between 87%-92% hydration, with
 725 the highest optima occurring at 33°C and when frogs were nearly fully hydrated at 98% (Figure
 726 3.4). Across 8, 15, and 24°C, frog performance declined past 10% of their maximum values after
 727 reaching 15-20% dehydration. However, at 33°C, this threshold is reached at 10% dehydration.
 728 Maximum performance is lowest at the coldest temperature (8°C: max performance = 0.468).

729 Performance was highest at the warmest temperature (33°C max performance = 0.825) but
730 declined rapidly below 90% hydration (Figure 3.4). Hydro performance curves showed
731 consistent shapes across temperatures, with performance remaining relatively stable until
732 hydration dropped below ~20%, after which all curves declined similarly (Figure 3.4). Individual
733 averaged curves are presented in Figure 3.3, illustrating variation in hydric performance across
734 frogs under different temperature treatments.

735

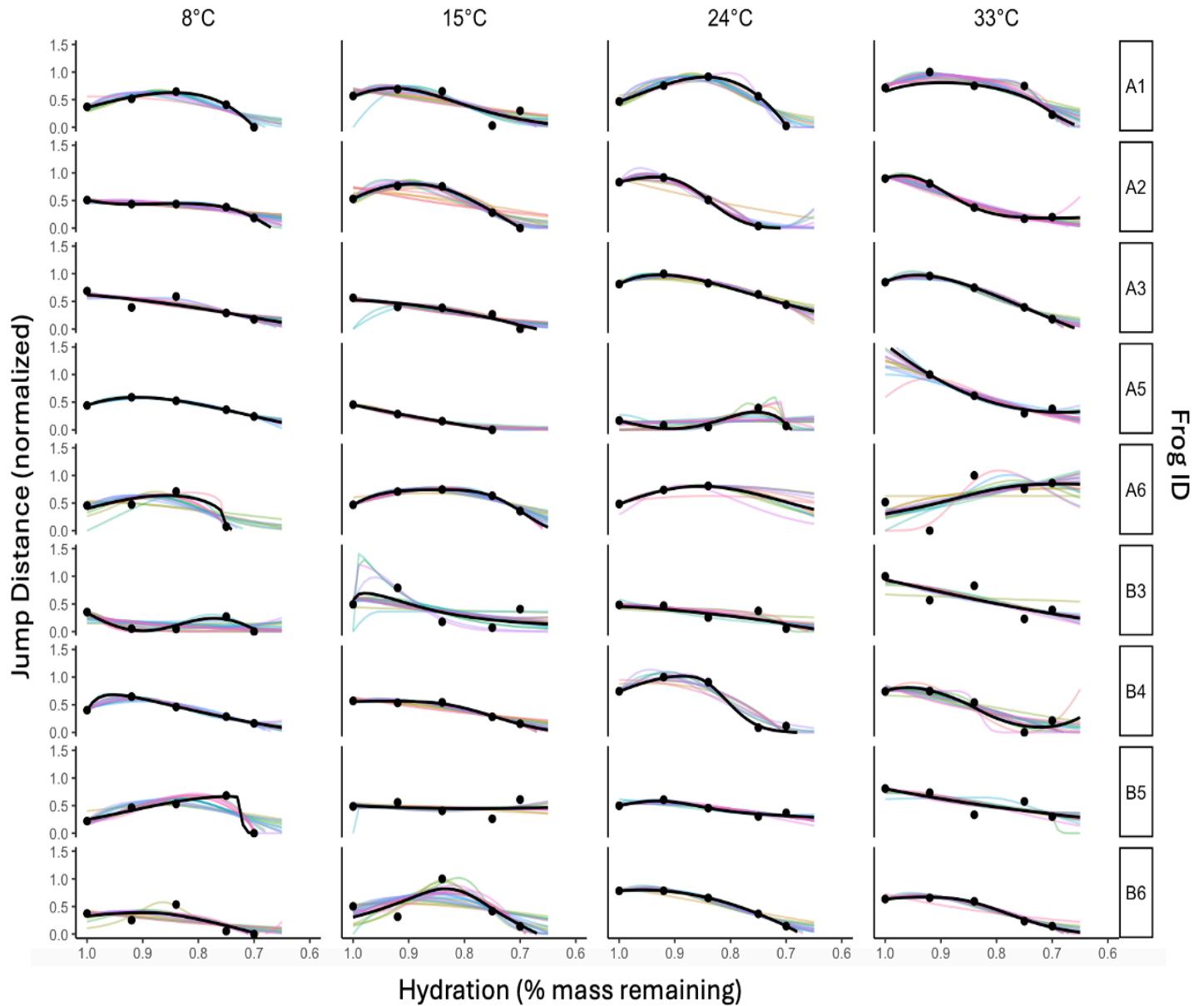
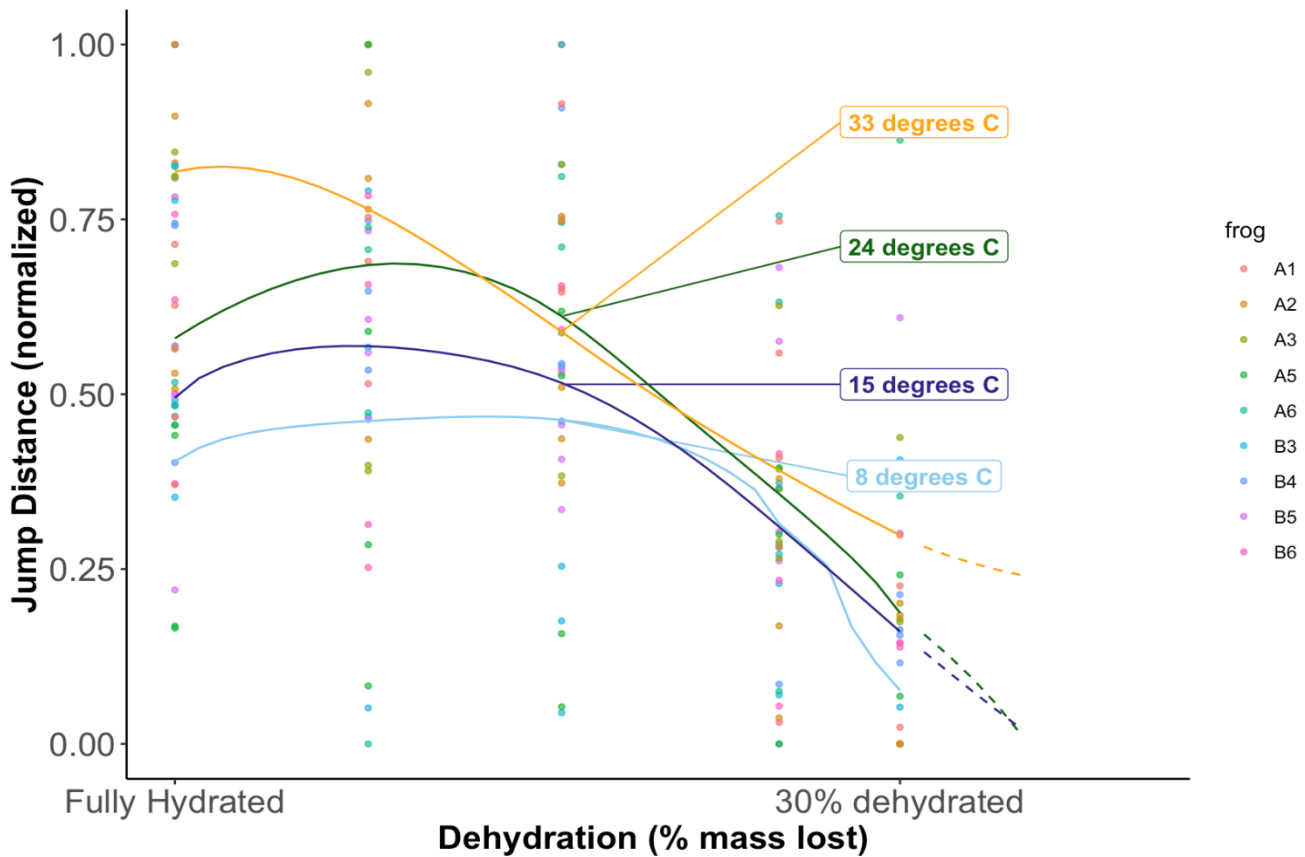


Figure 3.3 Individual weighted-average hydro-performance curves for all frogs and all temperature subsets. Faded coloured lines for each frog (each square) indicate single curves fit from model types (see appendix). Solid black lines for each plot are results from Akaike weighted average curves for each frog. Performance was measured as jump distance, normalized as the proportion of each frog's maximum jump distance across all temperatures. Points indicate these standardized distances for each frog at each hydration level (100, 92, 84, 75, 70 %) Hydration was measured as percentage of fully hydrated mass lost. Not all models were fit for each frog due to parameter (k) restrictions and final curve shapes for each model.

748



749

750

751 **Figure 3.4** Hydro performance curves at four temperatures. Data for trials was subset for water
 752 loss rates at each temperature (33°C, 24°C, 15°C, 8°C) across five dehydration levels (100%,
 753 92%, 84 %, 75%, 70%). Performance was measured as jump distance, normalized as the
 754 proportion of each frog's maximum jump distance across all temperatures. Hydration was
 755 measured as a percentage of fully hydrated mass lost. Curves are averaged from the weighted
 756 curves fit to each frog. Coloured points correspond to normalized maximum jump distance for
 757 frogs (A1, A2, A3, etc.)

758 3.3 Hydro-thermal Surface

759 To estimate the combined effects of temperature and hydration, I constructed a
 760 hydrothermal performance surface using the model-averaged curves above. This surface
 761 illustrates predicted jump performance across the range of conditions observed in the field
 762 (Figure 3.5).

763

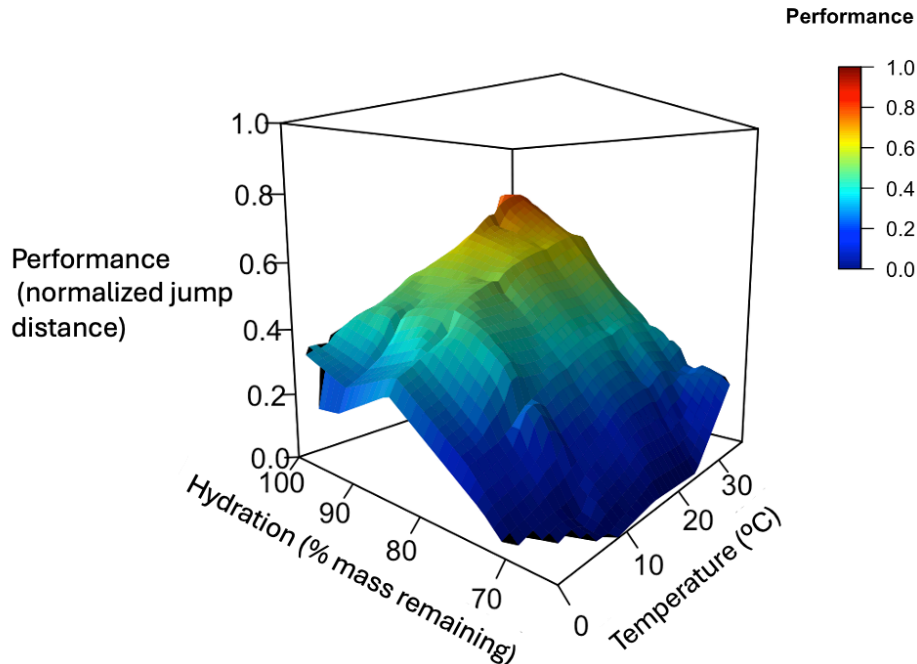


Figure 3.5 Hydrothermal performance surface generated from merging the univariate curves for hydro-performance and thermal performance curves and interpolating across a bilinear grid. Performance was measured as jump distance, normalized as the proportion of each frog's maximum jump distance across all temperatures. Hydration was measured as a percentage of fully hydrated mass lost. Temperature was measured from incubation temperatures in which frogs were held during trials.

3.4 H1 - Successional Differences in Macrohabitat

For the first hypothesis that forest and cut macrohabitats would be significantly different immediately post-cut and that these differences would persist 20-years post-cut, I tested interaction between macrohabitat (plot) and successional stage with additive effect of microhabitat (Table 3.1 and Table 3.2).

3.4.1 Daytime habitat and age interactions

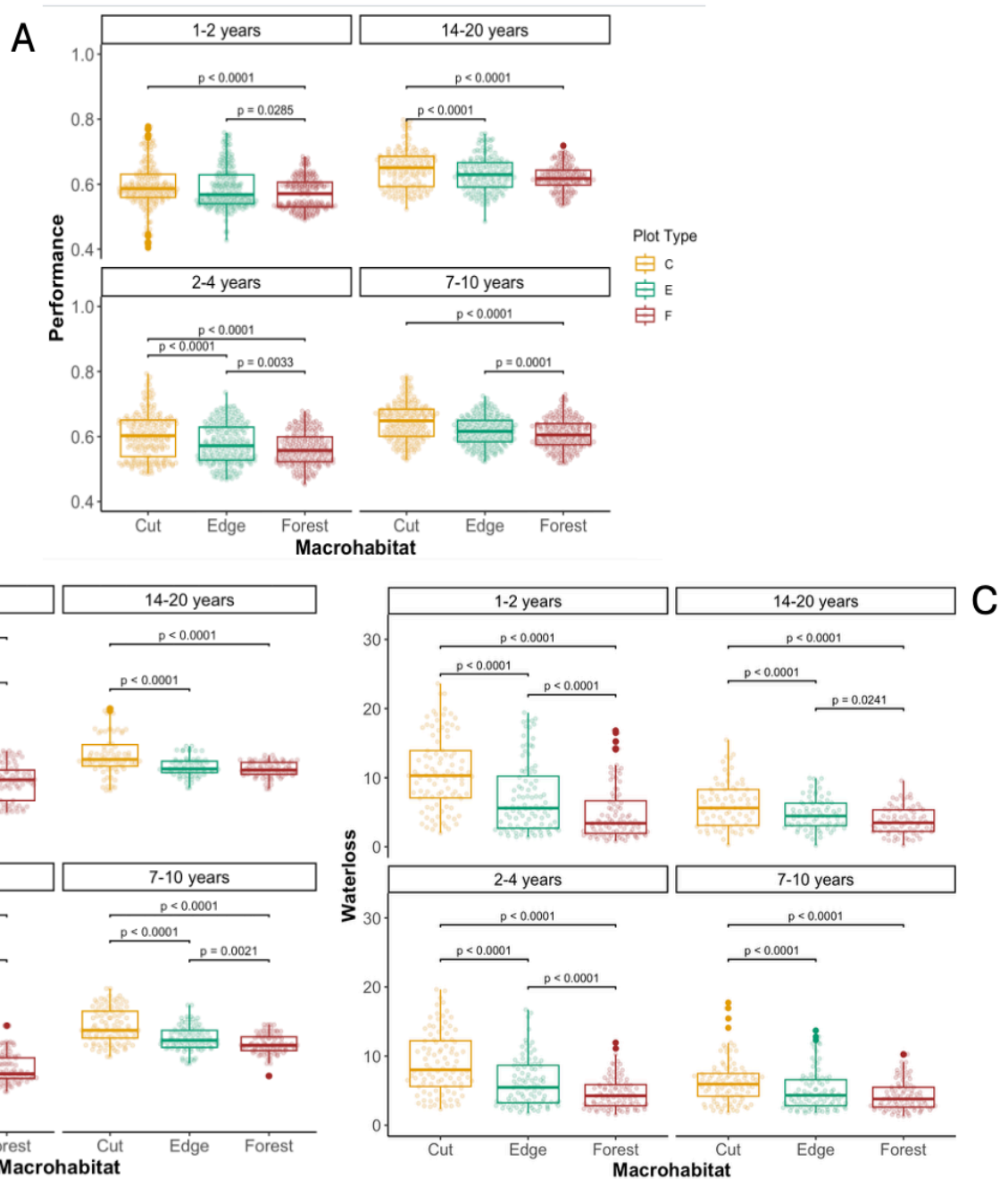
I found a significant main effect of microhabitat, as well as a significant interaction between plot type and age for all three response variables (temperature, water loss, performance fit for daytime data (Table 3.1).

Table 3.1 Hypothesis one model results for all three response variables during the day. Significant tests used marginal sums of squares for linear mixed effects models. Df indicates the numerator degrees of freedom. For the daytime water loss model, the denominator degrees of freedom (denDF) are: 22262 for plot type, microhabitat, and the plot type \times age interaction; and 12 for age. For both the daytime water loss and performance models, the denominator degrees of freedom (denDF) are: 2120 for plot type, microhabitat, and the plot type \times age interaction; and 11 for age. Denominator degrees of freedom for linear mixed effects models with lme were calculated using Kenward-Roger approximations. Significant p-values are in bold.

Response	Term	Df	F value	p-value
Temperature	plot type	2	18.52	<0.001
	age	3	1.72	0.215
	microhabitat	2	39.71	<0.001
	plot type x age	6	16.92	<0.001
Hydration	plot type	2	15.34	<0.001
	age	3	4.16	0.034
	microhabitat	2	452.55	<0.001
	plot type x age	6	27.35	<0.001
Performance	plot type	2	11.25	<0.001
	age	3	2.75	0.093
	microhabitat	2	207.78	<0.001
	plot type x age	6	3.65	0.001

Post-hoc tests for the interaction between macrohabitat and clearcut age indicated significant differences between habitats for all ages (Figure 3.6). For temperature effects, it was significantly warmer in cut plots than both forest and edge habitats, with the largest difference

794 between cut and forests 2-4 years old (mean difference = 1.715 °C, SE = 0.0973, $p < 0.001$). The
795 smallest difference was in cuts 14-20 years old (mean difference = 0.563 °C, SE = 0.0973, $p <$
796 0.001). Differences in water loss rates between cut and forests were significant across all ages,
797 with the greatest difference in 1–2-year-old sites (mean difference = 5.15 % weight lost, SE =
798 0.232, $p < 0.001$) and diminishing over time but remaining significant in our oldest successional
799 stage (14-20-years-old: mean difference = 1.492 % weight lost, SE = 0.269, $p < 0.001$).
800 Estimated performance differences were significant between cut and forests habitats across all
801 ages (Figure 3.6) with the smallest difference occurring in 14-20 year-old cuts (mean difference
802 = 0.024, SE = 0.0052, $p < 0.001$) and the largest differences occurring in our 7-10-year-old and
803 14-20-year-old successional plots (Figure 3.6).



805 **Figure 3.6** Daytime performance (A), temperature (B), and water loss (C) differences between
 806 macrohabitats (plot types: cut, edge, and forest). All results are averaged over microhabitat type.
 807 Results were generated from mixed effects models with fixed effects of macrohabitat, age,
 808 microhabitat, and an interaction term between macrohabitat x age. The model was fit with
 809 random effect of date nested within site. Points indicate all raw data. Bars indicate significant
 810 differences between pairs of macrohabitats with associated p-values (significant if $p < 0.05$).
 811 Comparisons were generated from estimated marginal means of difference between
 812 macrohabitats across age groups with Tukey adjusted p-values for multiple comparisons.

3.4.2 Overnight habitat and age interactions

I found a significant main effects of microhabitat, as well as a significant interaction between plot type and age for all three models fit for overnight data (Table 3.2).

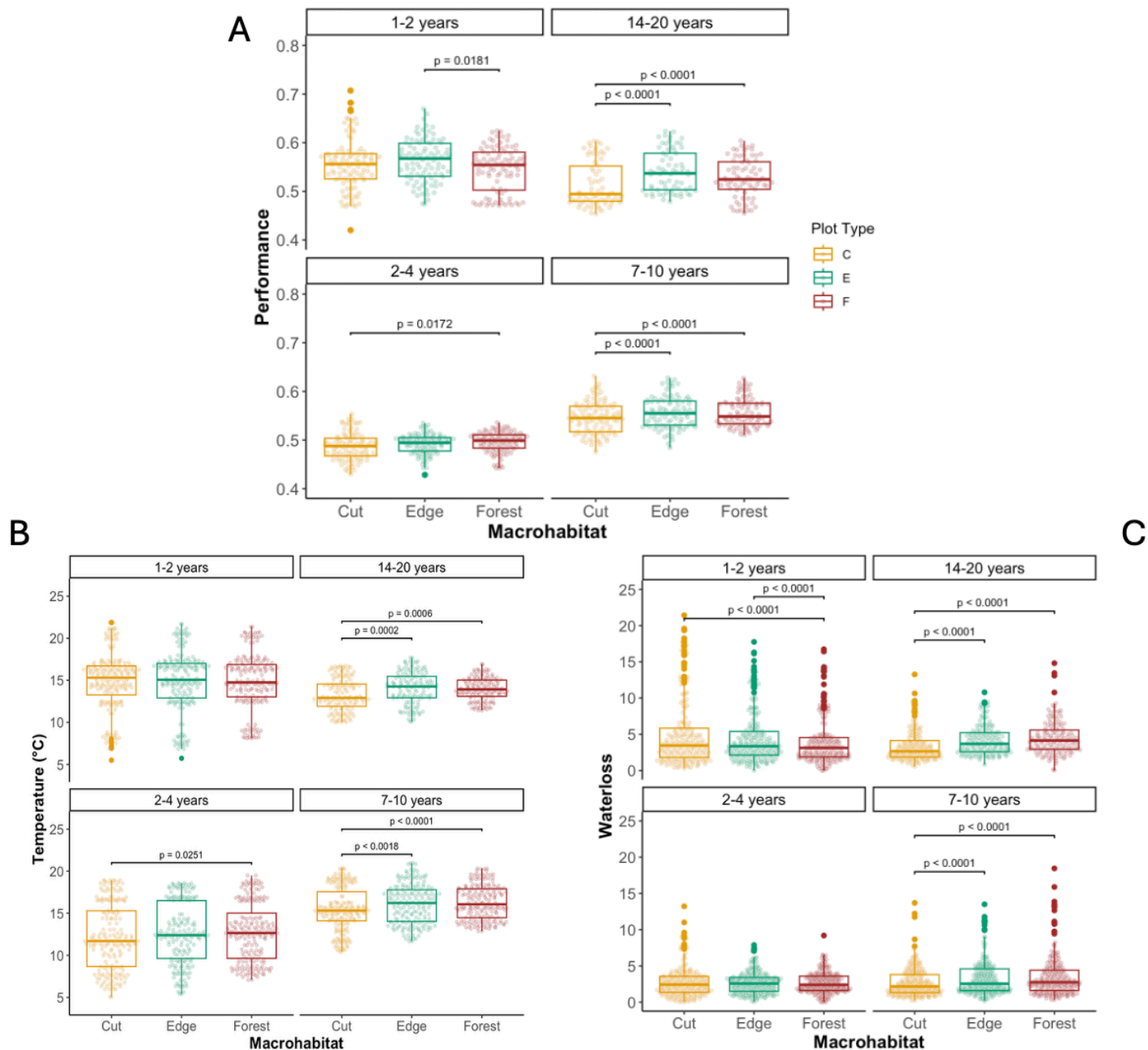
Table 3.2 Hypothesis one model results for all three response variables fit for overnight data. Left side table indicates models fit with temperature and performance ANOVA summaries generated using marginal type significance estimates and F-statistics. Right table is the model fit with hydration as a response with summaries generated using likelihood ratio tests with estimated chi squared distributions and model fit with as a glmm. For the overnight performance model, the denominator degrees of freedom (denDF) are: 2069 for plot type, microhabitat, and the plot type \times age interaction; and 11 for age. For the overnight temperature model, the denominator degrees of freedom (denDF) are: 2120 for plot type, microhabitat, and the plot type \times age interaction; and 11 for age. Denominator degrees of freedom for linear mixed effects models with lme were calculated using Kenward-Roger approximations. Significant effects are indicated by bold p-values.

Response	Term	Df	F value	p-value	Response	Term	Df	Chisq	p-value
Temperature	plot type	2	9.522	<0.001	Hydration	plot type	2	60.518	<0.001
	age	3	1.851	0.1917		age	3	2.974	0.396
	microhabitat	2	241.334	<0.001		microhabitat	2	1218.390	<0.001
	plot type x age	6	1.990	0.0643		plot type x age	6	99.297	<0.001
Performance	plot type	2	21.791	<0.001					
	age	3	1.349	0.3090					
	microhabitat	2	595.230	<0.001					
	plot type x age	6	7.413	<0.001					

Post-hoc tests for the interaction between macrohabitat (plot type) and clearcut age revealed significant differences in performance between habitats for all ages overnight (Figure 3.7 and Table 3.2). For temperature effects, it was significantly warmer in forest plots than both cut and edge habitats in all successional groups except in 1-2 year-since-cut groups, where no significant difference was found (mean difference = 0.213 °C, SE = 0.1647, $p < 0.399$). The

largest difference between cut and forests occurred in 7–10-year-old sites (mean difference = -0.783 °C, SE = 0.165, $p < 0.001$), where it was colder in clearcuts than in forests. Water loss was significantly higher in forests than in cuts (Figure 3.7) in almost all age groups except in 2–4 years-since-cut groups (mean difference = -0.117 % weight lost, SE = 0.0895, $p = 0.393$). Water loss was only significantly greater in 2–4-year-old cuts than in forests (mean difference = 0.5213 % weight lost, SE = 0.1070, $p < 0.001$). The largest difference between water loss in cut and forested habitats occurred in 14–20-year-old sites, wherein frogs lost more water in forests (mean difference = -0.737 % weight lost, SE = 0.1234, $p < 0.001$). Resulting performance was significantly higher in forests than in cuts across all ages except in 1–2-year-old sites (Figure 3.7). Performance differences between forest and cuts were largest in 7–10-year-old sites, when forests were significantly higher than cut plots (mean difference = -0.017, SE = 0.0027, $p < 0.001$).

Performance was higher in nearly all time-since-cut groups in forest compared to cut and macrohabitats. This difference was significant in all cut-ages except for 1–2-year-old cuts (Figure 3.7).



8.

850 **Figure 3.7** Overnight performance (A), temperature (B), and water loss (C) differences between
 851 macrohabitats (plot types). Results were generated from mixed effects models with fixed effects
 852 of macrohabitat, age, microhabitat, and an interaction term between macrohabitat x age. The
 853 model was fit with random effect of date nested within site. Large bold points indicate data
 854 outliers, and faded points indicate all raw data. Bars indicate significant differences between
 855 pairs of macrohabitats with associated p-values (significant if $p < 0.05$). Comparisons were
 856 generated from estimated marginal means of difference between plot types across age groups
 857 with Tukey adjusted p-values for multiple comparisons.

3.5 Microhabitat and postural differences across succession in macrohabitats

For the second hypothesis that overnight differences between cut and forests in uncovered microhabitats between would be more pronounced than differences in covered microhabitats during the day, I tested the interactions between macrohabitat (plot), successional stage, and microhabitat (Table 3.3 and Table 3.7).

3.5.1 Daytime microhabitat effects

Models fit for temperature during daytime periods revealed significant interactions between microhabitat and age and plot type with age, but no significant interaction of plot type with microhabitat (Table 3.3). Models for both hydration and temperature revealed significant interactions between all response variables of plot type, microhabitat, and age (Table 3.3).

Table 3.3 Models testing hypothesis two and interactions of plot and age with microhabitat. All estimates were generated from type III significance tests using denominator degrees of freedom (ddf) F tests for p-values. Significant estimates are indicated by bolded p-values. “Df” indicates numerator degrees of freedom for each parameter.

Response	Term	Df	Chisq	p-value
Temperature	plot type	2	53.74	<0.001
	microhabitat	2	4.12	0.127
	plot type x age	9	164.96	<0.001
	plot type x microhabitat	4	6.67	0.154
	age x microhabitat	6	133.50	<0.001
Hydration	plot type	2	8.93	0.012
	microhabitat	2	345.91	<0.001
	plot type x age	9	206.90	<0.001
	plot type x microhabitat	4	110.45	<0.001
	age x microhabitat	6	22.64	0.001
Performance	plot type	2	22.35	<0.001
	microhabitat	2	200.99	<0.001
	plot type x age	9	29.64	0.001
	plot type x microhabitat	4	10.88	0.028
	age x microhabitat	6	269.08	<0.001

Post hoc tests for comparisons of temperatures in microhabitats between plots during the day (e.g. uncovered cut vs. uncovered forest) demonstrated significantly higher average temperatures in all cut habitats than in forested habitats in both covered and uncovered microhabitats for all ages (Table 3.4). Differences between temperatures in cut and forests were larger for uncovered microhabitats, and the largest difference by year occurs in 7-10-year-old sites for both covered (mean difference = 1.417 °C, SE = 0.086, $p < 0.001$) and uncovered (mean difference = 1.4363 °C, SE = 0.084, $p < 0.001$) microhabitats (Figure 3.8). There were no significant differences between microhabitats within habitat types except between covered and uncovered forest microhabitats in 7–10-year-old sites (Table 3.4).

896 **Table 3.4** Daytime estimated marginal means for microhabitat comparisons within (left) and
897 between (right) macrohabitat types for temperature. Results were generated from Tukey corrected
898 least squares means on the mixed effects models for temperature and hypothesis two. “Age”
899 indicates the time-since-cut group, “contrast” indicates the microhabitat and plot contrasts, with
900 “estimate” as the difference between the estimated marginal means for each term in the pairing
901 going from left to right. Significant contrasts ($p < 0.05$) are indicated in bold.

Contrasts within macrohabitats					Contrasts between macrohabitats				
Age	Contrast	Estimate	SE	p-value	Age	Contrast	Estimate	SE	p-value
1-2 years	tree covered C - tree uncovered C	0.060	0.079	0.998	1-2 years	tree covered C - tree covered F	0.854	0.080	< 0.001
	tree covered F - tree uncovered F	0.079	0.073	0.977		tree uncovered C - tree uncovered F	0.873	0.080	< 0.001
2-4 years	tree covered C - tree uncovered C	0.112	0.081	0.904	2-4 years	tree covered C - tree covered F	1.158	0.082	< 0.001
	tree covered F - tree uncovered F	0.131	0.072	0.670		tree uncovered C - tree uncovered F	1.178	0.079	< 0.001
7-10 years	tree covered C - tree uncovered C	0.236	0.086	0.132	7-10 years	tree covered C - tree covered F	1.417	0.086	< 0.001
	tree covered F - tree uncovered F	0.255	0.076	0.021		tree uncovered C - tree uncovered F	1.436	0.084	< 0.001
14-20 years	tree covered C - tree uncovered C	0.110	0.080	0.911	14-20 years	tree covered C - tree covered F	0.393	0.081	< 0.001
	tree covered F - tree uncovered F	0.129	0.075	0.732		tree uncovered C - tree uncovered F	0.412	0.079	< 0.001

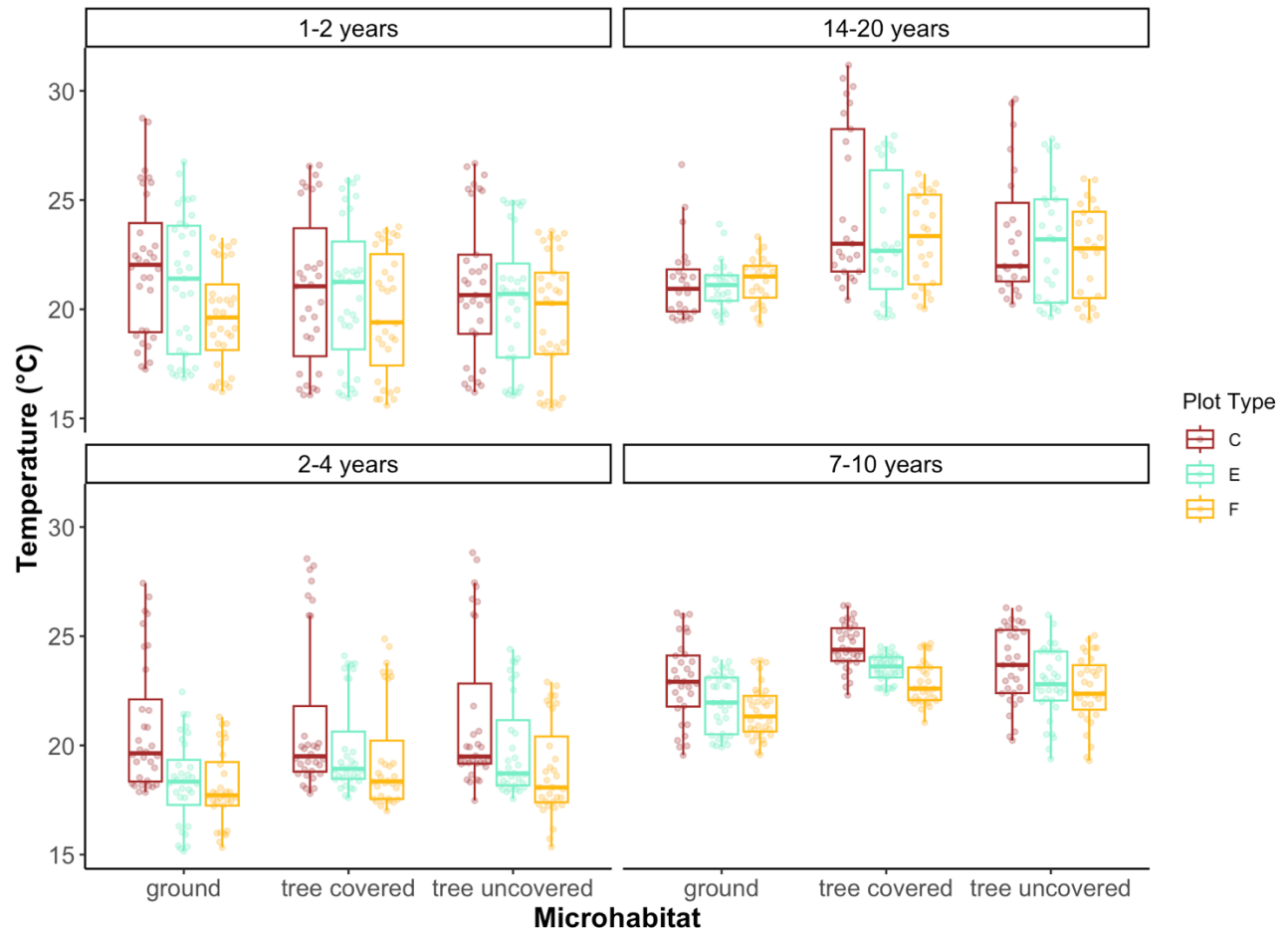


Figure 3.8 Daytime temperature differences between macrohabitats (plot types: cut, edge, and forest). Results were generated from mixed effects models with fixed effects of plot type (macrohabitat), age, microhabitat, and interactions between all three. The model was fit with random effect of date nested within site. Points indicate all raw data. Comparisons were generated from estimated marginal means of difference between microhabitat and macrohabitat interactions across age groups with Tukey adjusted p-values for multiple comparisons.

Post hoc tests for comparisons of water loss rates in microhabitats between plots during the day (e.g. uncovered cut vs. uncovered forest) demonstrated significantly higher water loss rates in cut habitats than in forested habitats in both covered and uncovered microhabitats for all ages (Table 3.5). Differences in water loss rates between cut and forests were larger for uncovered microhabitats, and the largest difference by year occurs in 1-2-year-old sites for both covered (mean difference = 4.9029 % weight lost, SE = 0.242, $p < 0.001$) and uncovered (mean difference = 5.3126 % weight lost, SE = 0.084, $p < 0.001$) microhabitats. Differences between

917 microhabitats diminished throughout succession but were still significant in our earliest
918 successional stage treatment (14-20-years; Table 3.5). Water loss was significantly higher in
919 uncovered microhabitats within habitat types (e.g. covered cut vs uncovered cut) for all age
920 groups in both forest and clearcut habitats (Table 3.5). Differences between water loss rates in
921 microhabitats was larger in cuts than in forests for all age groups, and differences were greatest
922 in the most recent successional stage (1-2-year-old: mean difference = -1.67 % weight lost, SE =
923 0.261, $p < 0.001$), with uncovered replicas losing more water (Figure 3.9).

924 **Table 3.5** Daytime estimated marginal means for contrasts in water loss rates for microhabitat
925 comparisons **within** (left) and **between** (right) macrohabitat (plot) types. Results were generated
926 from Tukey corrected least squares means on the mixed effects models for water loss in
927 hypothesis two. “Age” indicates the time-since-cut grouping, “contrast” indicates the
928 microhabitat and plot contrasts and direction, with “estimate” as the difference between the
929 estimated marginal means for each term in the pairing. Significant contrasts ($p < 0.05$) are
930 indicated in bold.

Contrasts within macrohabitats					Contrasts between macrohabitats				
Age	Contrast	Estimate	SE	p-value	Age	Contrast	Estimate	SE	p-value
1-2 years	tree covered C - tree uncovered C	-1.670	0.261	< 0.001	1-2 years	tree covered C - tree covered F	4.903	0.242	< 0.001
	tree covered F - tree uncovered F	-1.260	0.218	< 0.001		tree uncovered C - tree uncovered F	5.313	0.254	< 0.001
2-4 years	tree covered C - tree uncovered C	-1.314	0.251	< 0.001	2-4 years	tree covered C - tree covered F	3.300	0.234	< 0.001
	tree covered F - tree uncovered F	-0.905	0.201	< 0.001		tree uncovered C - tree uncovered F	3.710	0.244	< 0.001
7-10 years	tree covered C - tree uncovered C	-1.517	0.221	< 0.001	7-10 years	tree covered C - tree covered F	1.901	0.209	< 0.001
	tree covered F - tree uncovered F	-1.108	0.201	< 0.001		tree uncovered C - tree uncovered F	2.311	0.213	< 0.001
14-20 years	tree covered C - tree uncovered C	-1.744	0.249	< 0.001	14-20 years	tree covered C - tree covered F	1.347	0.236	< 0.001
	tree covered F - tree uncovered F	-1.335	0.231	< 0.001		tree uncovered C - tree uncovered F	1.757	0.244	< 0.001

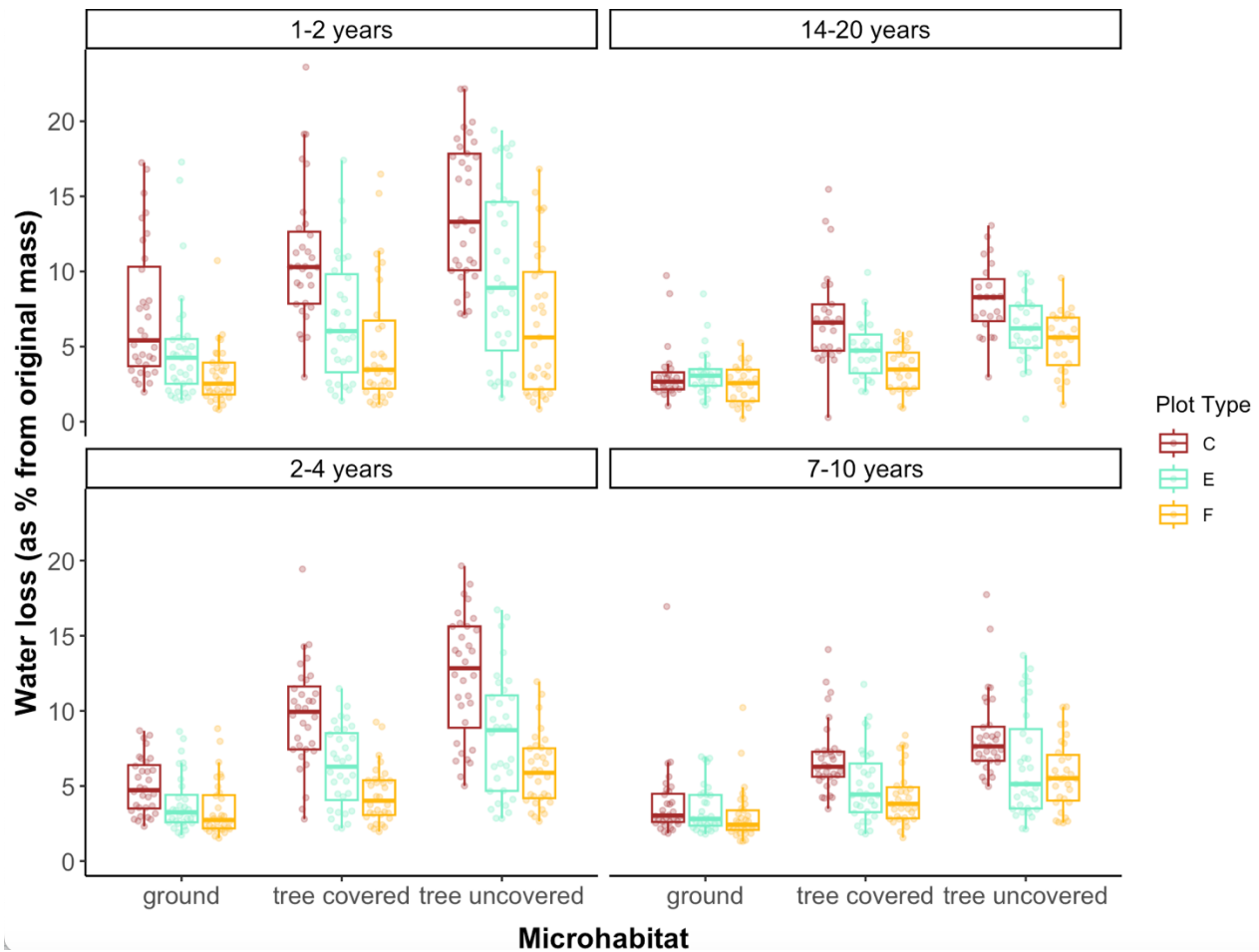


Figure 3.9 Daytime water loss differences between macrohabitats (plot types: cut, edge, and forest). Results were generated from mixed effects models with fixed effects of plot type (macrohabitat), age, microhabitat, and interactions between all three. The model was fit with random effect of date nested within site. Points indicate all raw data. Comparisons were generated from estimated marginal means of difference between microhabitat and macrohabitat interactions across age groups with Tukey adjusted p-values for multiple comparisons.

Post-hoc tests for comparisons of estimated performance in microhabitats between plots during the day (e.g. uncovered cut vs. uncovered forest) indicated significantly higher performance in cut habitats than in forested habitats in both covered and uncovered microhabitats for all ages (Table 3.6). There were no significant differences for performance in cuts between uncovered and covered microhabitats except in the 14-20-year-old cuts (Table 3.6). Between clearcuts and forests, performance was higher in both covered cut and uncovered cuts relative to covered forest and uncovered forest microhabitats (Table 3.6), and the largest

946 difference occurred in 2-4-year-old sites for both covered (mean difference = 0.0451, SE =
947 0.003, $p < 0.001$) and uncovered (mean difference = 0.0367, SE = 0.003, $p < 0.001$)
948 microhabitats. Differences between microhabitats diminished throughout succession but were
949 still significant in our earliest successional stage treatment (14-20-years-old; Table 3.6).

950 Comparisons of estimated performance in uncovered vs. covered microhabitats within
951 plots during the day (e.g. uncovered cut vs. covered cut) indicated significant differences in both
952 forest and clearcut habitats only in our earliest age treatment (14-20-years-old; Table 3.6; Figure
953 3.10). Differences in performance were greater in forests between covered and uncovered
954 microhabitats for the two earliest time-since-cut groups (14-20 years and 7-10 years), with
955 higher estimated performance in uncovered forest microhabitat (Table 3.6; Figure 3.10). There
956 were no significant differences between microhabitats within habitat types 2-4 or 1-2-year-old
957 sites (Table 3.6).

969 **Table 3.6.** Daytime estimated means for contrasts in performance for microhabitat comparisons
970 **within** (left) and **between** (right) macrohabitat (plot) types. Results were generated from Tukey
971 corrected least squares means on the mixed effects models for performance in hypothesis two.
972 “Age” indicates the time-since-cut group, “contrast” indicates the microhabitat and plot contrasts,
973 with “estimate” as the difference between the means for each term in the pairing. Significant
974 contrasts ($p < 0.05$) are indicated in bold.

Contrasts within macrohabitats					Contrasts between macrohabitats				
Age	Contrast	Estimate	SE	p-value	Age	Contrast	Estimate	SE	p-value
1-2 years	tree covered C - tree uncovered C	0.0047	0.004	0.955	1-2 years	tree covered C - tree covered F	0.0405	0.004	< 0.001
	tree covered F - tree uncovered F	-0.0037	0.003	0.965		tree uncovered C - tree uncovered F	0.0321	0.004	< 0.001
2-4 years	tree covered C - tree uncovered C	0.0004	0.004	1.000	2-4 years	tree covered C - tree covered F	0.0451	0.003	< 0.001
	tree covered F - tree uncovered F	-0.0080	0.003	0.204		tree uncovered C - tree uncovered F	0.0367	0.003	< 0.001
7-10 years	tree covered C - tree uncovered C	-0.0106	0.004	0.071	7-10 years	tree covered C - tree covered F	0.0427	0.003	< 0.001
	tree covered F - tree uncovered F	-0.0190	0.003	< 0.001		tree uncovered C - tree uncovered F	0.0343	0.004	< 0.001
14-20 years	tree covered C - tree uncovered C	-0.0167	0.004	< 0.001	14-20 years	tree covered C - tree covered F	0.0259	0.004	< 0.001
	tree covered F - tree uncovered F	-0.0251	0.004	< 0.001		tree uncovered C - tree uncovered F	0.0176	0.004	< 0.001

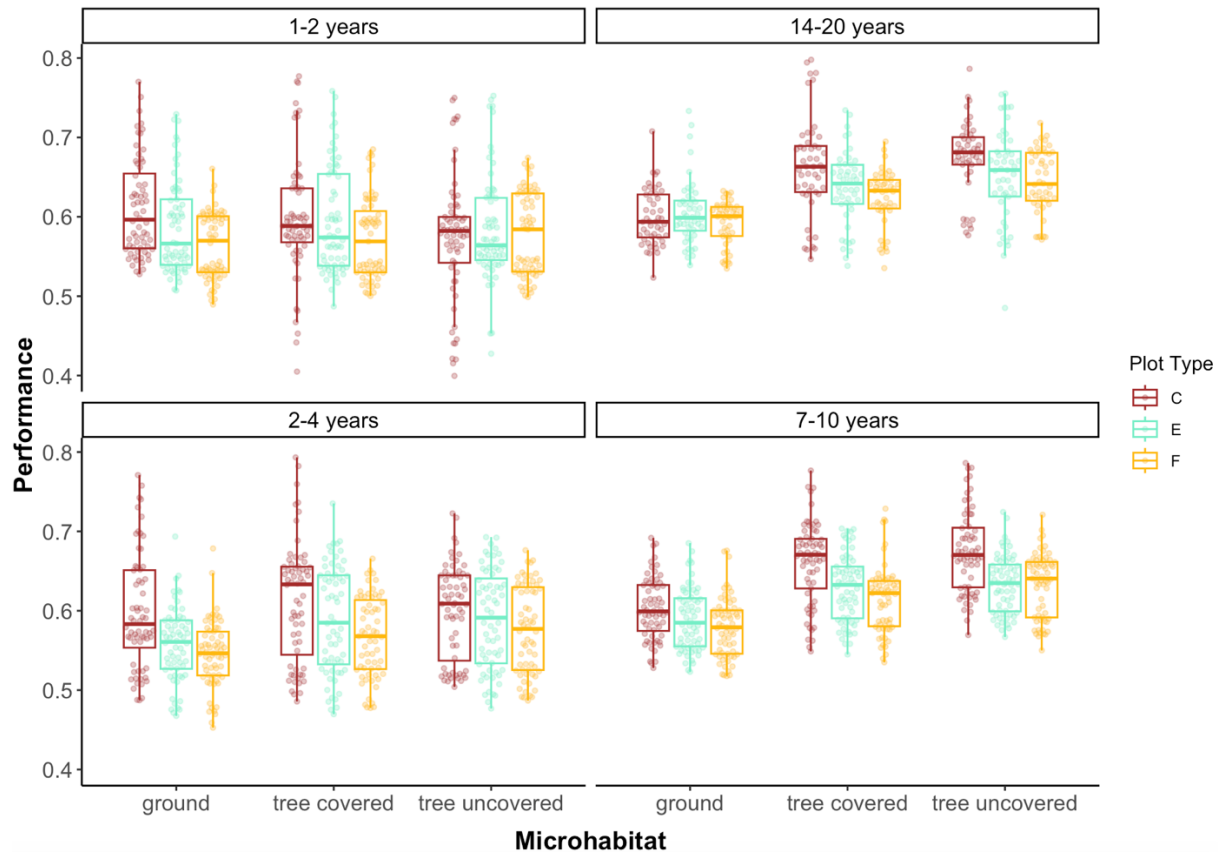


Figure 3.10 Daytime performance differences between macrohabitats (plot types: cut, edge, and forest). Results were generated from mixed effects models with fixed effects of plot type (macrohabitat), age, microhabitat, and interactions between all three. The model was fit with random effect of date nested within site. Points indicate all raw data. Comparisons were generated from estimated marginal means of difference between microhabitat and macrohabitat interactions across age groups with Tukey adjusted p-values for multiple comparisons.

3.5.2 Overnight microhabitat effects

Models fit for temperature during overnight periods revealed significant interactions between microhabitat and age and plot type with age, but no significant interaction of plot type with microhabitat (Table 3.7). Models for hydration indicated significant interactions between all response variables plot type, microhabitat, and age (Table 3.7). Models for performance indicated significant interactions between age with plot, and plot with microhabitat, but no significant interaction between age and microhabitat (Table 3.7).

Table 3.7 Models testing hypothesis two and interactions of plot and age with microhabitat for overnight data. “Df” indicates numerator degrees of freedom for each level. All estimates were generated from likelihood ratio tests and Chi-squared statistics for p-values. Significant contrasts are indicated by bold p-values.

Response	Term	Df	Chisq	p-value
Temperature	plot type	2	18.19	<0.001
	microhabitat	2	35.31	<0.001
	plot type x age	9	29.56	0.001
	plot type x microhabitat	4	7.72	0.102
	age x microhabitat	6	31.30	<0.001
Hydration	plot type	2	55.75	<0.001
	microhabitat	2	286.97	<0.001
	plot type x age	9	97.76	<0.001
	plot type x microhabitat	4	35.48	<0.001
	age x microhabitat	6	33.11	<0.001
Performance	plot type	2	81.63	<0.001
	microhabitat	2	177.68	<0.001
	plot type x age	9	94.78	<0.001
	plot type x microhabitat	4	16.90	0.002
	age x microhabitat	6	11.47	0.075

Post hoc tests for comparisons of microhabitat temperatures between plots overnight (e.g. uncovered cut vs. uncovered forest) demonstrated significantly lower average temperatures in cut habitats than in forested habitats in both covered and uncovered microhabitats for all ages except in covered microhabitats in 1-2-year-old sites (mean difference = -0.255 °C, SE = 0.099, $p = 0.194$; Table 3.8; Figure 3.11). Differences between temperatures in cut and forests were larger for uncovered microhabitats, and the largest difference occurred in 7-10-year-old sites for both covered (mean difference = -0.713 °C, SE = 0.119, $p < 0.001$) and uncovered (mean difference = -0.7962 °C, SE = 0.116, $p < 0.001$) microhabitats (Table 3.8; Figure 3.11). In microhabitats

1004 within plot types (e.g. covered cut vs. uncovered cut), covered microhabitats were significantly
 1005 warmer than uncovered ones (Table 3.8). Temperature differences between microhabitats within
 1006 each macrohabitat type were significantly larger in cuts than in forests and were largest in 14-20
 1007 and 2-4-year-old sites (Table 3.8; Figure 3.11).

1008 **Table 3.8** Overnight estimated marginal means for contrasts in temperature for microhabitat
 1009 comparisons **within** (left) and **between** (right) macrohabitat (plot) types. Results were generated
 1010 from Tukey corrected least squares means on the mixed effects models for temperature in
 1011 hypothesis two. “Age” indicates the time-since-cut group, “contrast” indicates the microhabitat
 1012 and plot contrasts, with “estimate” as the difference between the estimated means for each term
 1013 in the pairing going from left to right. Significant contrasts ($p < 0.05$) are indicated in bold.

Contrasts within macrohabitats					Contrasts between macrohabitats				
Age	Contrast	Estimate	SE	p-value	Age	Contrast	Estimate	SE	p-value
1-2 years	tree covered C - tree uncovered C	0.4174	0.100	0.001	1-2 years	tree covered C - tree covered F	-0.255	0.099	0.194
	tree covered F - tree uncovered F	0.3342	0.092	0.009		tree uncovered C - tree uncovered F	-0.338	0.097	0.014
2-4 years	tree covered C - tree uncovered C	0.5729	0.106	<0.001	2-4 years	tree covered C - tree covered F	-0.553	0.103	<0.001
	tree covered F - tree uncovered F	0.4897	0.094	<0.001		tree uncovered C - tree uncovered F	-0.636	0.105	<0.001
7-10 years	tree covered C - tree uncovered C	0.4784	0.116	0.001	7-10 years	tree covered C - tree covered F	-0.713	0.119	<0.001
	tree covered F - tree uncovered F	0.3952	0.102	0.003		tree uncovered C - tree uncovered F	-0.796	0.116	<0.001
14-20 years	tree covered C - tree uncovered C	0.5636	0.107	<0.001	14-20 years	tree covered C - tree covered F	-0.484	0.106	<0.001
	tree covered F - tree uncovered F	0.4805	0.099	<0.001		tree uncovered C - tree uncovered F	-0.567	0.104	<0.001

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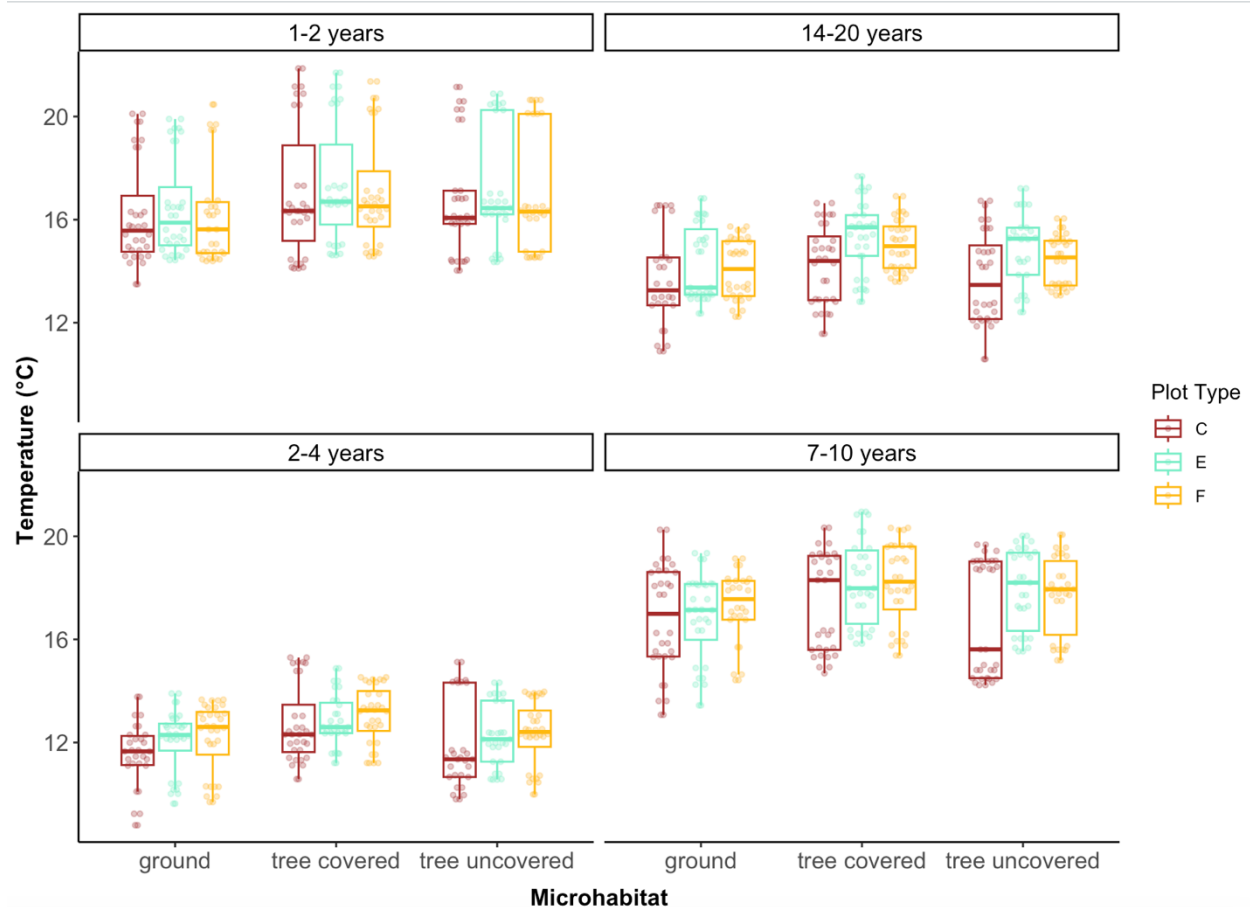


Figure 3.11 Overnight temperature differences between macrohabitats (plot types: cut, edge, and forest). Results were generated from mixed effects models with fixed effects of plot type (macrohabitat), age, microhabitat, and interactions between all three. The model was fit with random effect of date nested within site. Points indicate all raw data. Comparisons were generated from estimated marginal means of difference between microhabitat and macrohabitat interactions across age groups with Tukey adjusted p-values for multiple comparisons.

Post hoc tests for comparisons of water loss rates in microhabitats between plots overnight (e.g. uncovered cut vs. uncovered forest) indicated significantly higher water loss rates in uncovered forested habitats than in uncovered clearcut habitats for all ages except 1-2-year-old sites (Table 3.9; Figure 3.12). Water loss rates in covered cut microhabitats were significantly higher than covered forest microhabitats only in 1-2-year-old-sites (mean difference = 0.941 % weight lost, SE = 0.13, $p < 0.001$; Table 3.9), but were not significant for all other time-since-cut groups. Differences between microhabitats increased throughout succession, with

1032 larger differences clearcut and forest microhabitats overnight in our earliest successional stage
 1033 (14-20-year-old sites; Table 3.9).
 1034 Within macrohabitat types, water loss was significantly higher in uncovered
 1035 microhabitats (e.g. covered cut vs uncovered cut) for all age groups in only forested habitats but
 1036 were not significant for clearcuts (Table 3.9; Figure 3.12). Differences between covered and
 1037 uncovered forest microhabitats were greatest in our earliest successional group, where water loss
 1038 was higher in uncovered microhabitats (mean difference = -0.8828 % weight lost, SE = 0.148, p
 1039 < 0.001; Table 3.9; Figure 3.12).

1040 **Table 3.9.** Overnight estimated marginal means for contrasts in water loss for microhabitat
 1041 comparisons **within** (left) and **between** (right) macrohabitat (plot) types. Results were generated
 1042 from Tukey corrected least squares means on the mixed effects models for water loss in
 1043 hypothesis two. “Age” indicates the time-since-cut group, “contrast” indicates the microhabitat
 1044 and plot contrasts, with “estimate” as the difference between the estimated means for each term
 1045 in the pairing reading from left to right in the contrast column. Significant contrasts (p < 0.05)
 1046 are indicated in bold.

Contrasts within macrohabitats					Contrasts between macrohabitats				
Age	Contrast	Estimate	SE	p-value	Age	Contrast	Estimate	SE	p-value
1-2 years	tree covered C - tree uncovered C	0.1196	0.140	0.995	1-2 years	tree covered C - tree covered F	0.9417	0.130	< 0.001
	tree covered F - tree uncovered F	-0.6283	0.129	< 0.001		tree uncovered C - tree uncovered F	0.1938	0.132	0.868
2-4 years	tree covered C - tree uncovered C	0.2629	0.122	0.438	2-4 years	tree covered C - tree covered F	0.3152	0.117	0.149
	tree covered F - tree uncovered F	-0.4850	0.115	< 0.001		tree uncovered C - tree uncovered F	-0.4327	0.116	0.006
7-10 years	tree covered C - tree uncovered C	0.2156	0.125	0.734	7-10 years	tree covered C - tree covered F	-0.1165	0.123	0.990
	tree covered F - tree uncovered F	-0.5323	0.124	< 0.001		tree uncovered C - tree uncovered F	-0.8644	0.125	< 0.001
14-20 years	tree covered C - tree uncovered C	-0.1349	0.153	0.994	14-20 years	tree covered C - tree covered F	-0.2940	0.149	0.566
	tree covered F - tree uncovered F	-0.8828	0.148	< 0.001		tree uncovered C - tree uncovered F	-1.0419	0.146	< 0.001

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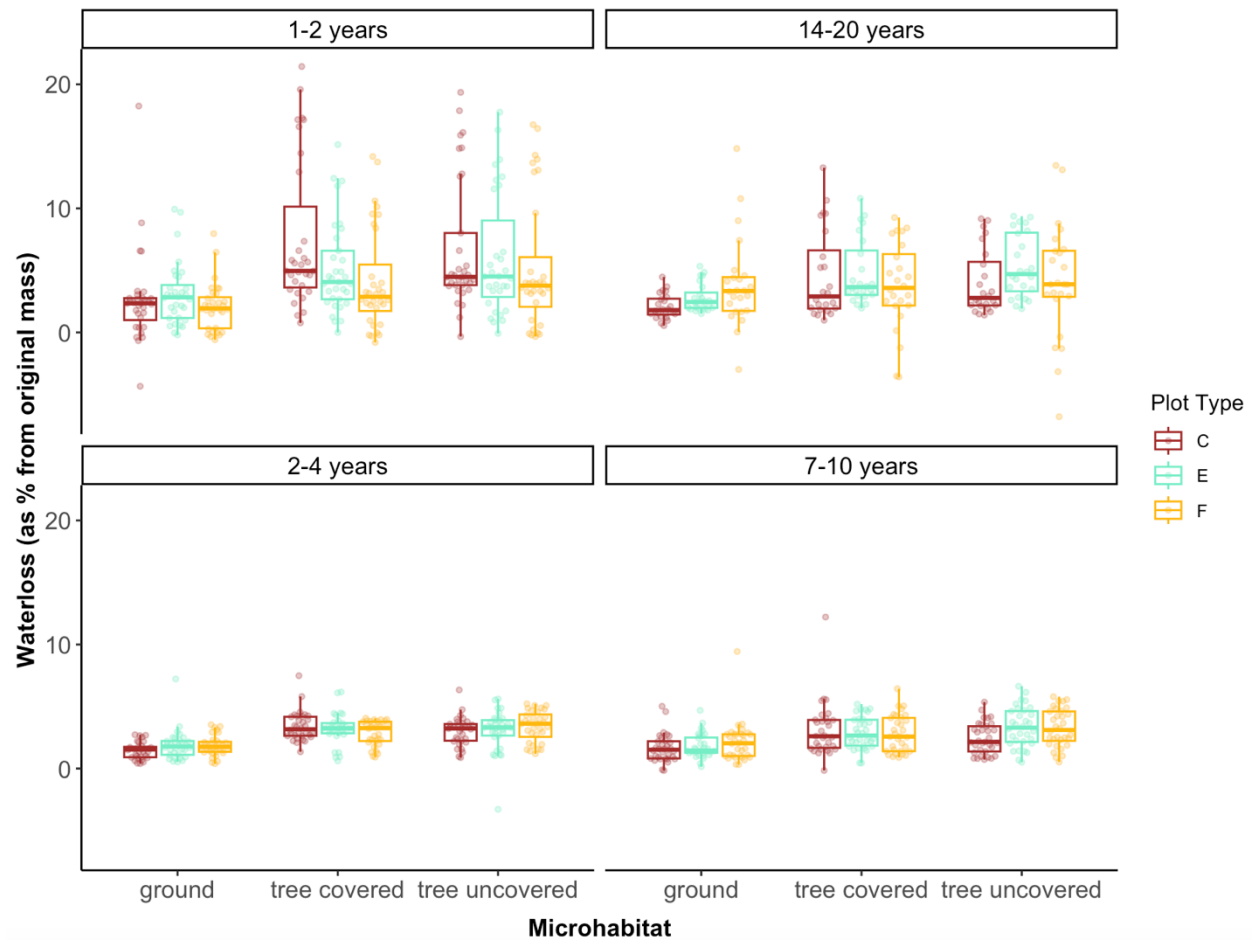


Figure 3.12 Overnight water loss differences between macrohabitats (plot types: cut, edge, and forest). Results were generated from mixed effects models with fixed effects of plot type (macrohabitat), age, microhabitat, and interactions between all three. The model was fit with random effect of date nested within site. Points indicate all raw data. Comparisons were generated from estimated marginal means of difference between microhabitat and macrohabitat interactions across age groups with Tukey adjusted p-values for multiple comparisons.

Post-hoc tests for comparisons of estimated performance in microhabitats between plots overnight (e.g. uncovered cut vs. uncovered forest) indicated significantly higher performance in forested than in clearcut habitats in both covered and uncovered microhabitats for all ages except 2-4 and 1-2-year-old cuts (Table 3.10). Performance was significantly highest in uncovered forest habitats compared with uncovered 7-10-year-old cuts (mean difference = -0.176, SE =

0.002, $p < 0.001$; Table 3.10). Performance was only higher in cuts at covered microhabitats 1-2 years old(mean difference = 0.0085, SE = 0.002, $p = 0.007$; Table 3.10)

Within microhabitats, performance was higher in covered microhabitats than in uncovered microhabitats for both cuts and forests (Table 3.10). Differences between microhabitats increased throughout succession but were still significant in our earliest successional stage treatment (14-20 years: Table 3.10; Figure 3.13). Differences in performance were only significant between microhabitats in clearcut habitats but not in forest habitats (Table 3.10; Figure 3.13). The largest difference between covered cuts and uncovered cut microhabitats occurred in 14-20-year-old cuts (mean difference = 0.0127, SE = 0.003, $p < 0.001$; Table 3.10).

Table 3.10 Estimated marginal means for contrasts in performance for microhabitat comparisons between habitat types overnight. Results were generated from Tukey corrected least squares means on the mixed effects models for performance in hypothesis two. “Age” indicates the successional group, “contrast” indicates the microhabitat and plot contrasts, with “estimate” as the difference between the estimated means for each term in the pairing. Significant contrasts ($p < 0.05$) are indicated in bold.

Contrasts within macrohabitats					Contrasts between macrohabitats				
Age	Contrast	Estimate	SE	p-value	Age	Contrast	Estimate	SE	p-value
1-2 years	tree covered C - tree uncovered C	0.008	0.002	0.019	1-2 years	tree covered C - tree covered F	0.008	0.002	0.008
	tree covered F - tree uncovered F	0.001	0.002	0.999		tree uncovered C - tree uncovered F	0.002	0.002	0.997
2-4 years	tree covered C - tree uncovered C	0.011	0.002	< 0.001	2-4 years	tree covered C - tree covered F	-0.004	0.002	0.754
	tree covered F - tree uncovered F	0.005	0.002	0.450		tree uncovered C - tree uncovered F	-0.010	0.002	< 0.001
7-10 years	tree covered C - tree uncovered C	0.008	0.002	0.029	7-10 years	tree covered C - tree covered F	-0.012	0.002	< 0.001
	tree covered F - tree uncovered F	0.001	0.002	1.000		tree uncovered C - tree uncovered F	-0.018	0.002	< 0.001
14-20 years	tree covered C - tree uncovered C	0.012	0.003	< 0.001	14-20 years	tree covered C - tree covered F	-0.009	0.003	0.007
	tree covered F - tree uncovered F	0.006	0.002	0.340		tree uncovered C - tree uncovered F	-0.016	0.003	< 0.001

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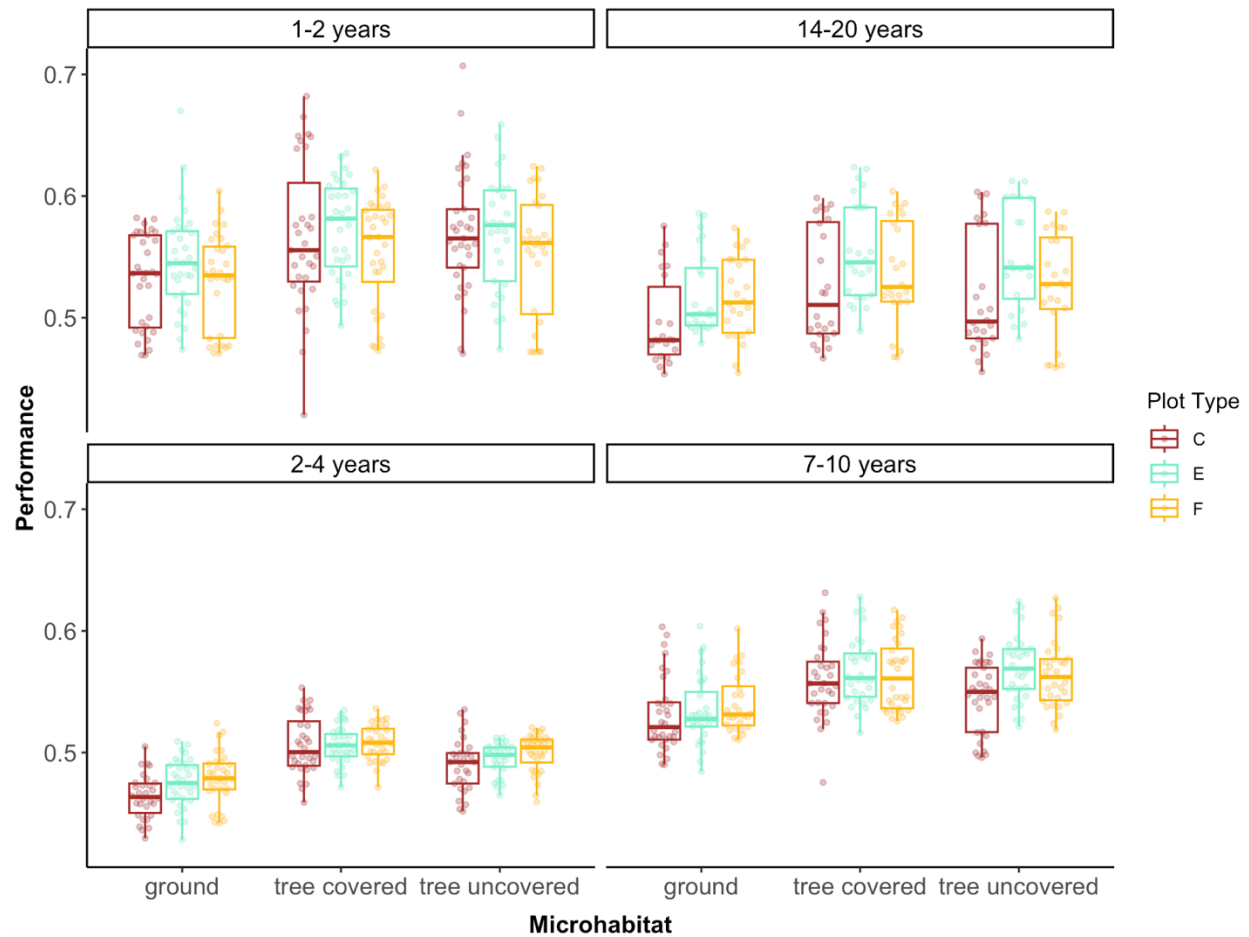


Figure 3.13 Overnight performance differences between macrohabitats (plot types: cut, edge, and forest). Results were generated from mixed effects models with fixed effects of plot type (macrohabitat), age, microhabitat, and interactions between all three. The model was fit with random effect of date nested within site. Points indicate all raw data. Comparisons were generated from estimated marginal means of difference between microhabitat and macrohabitat interactions across age groups with Tukey adjusted p-values for multiple comparisons.

3.5.3 Postural Effects

During the day, upright frogs had significantly higher performance than flat frogs for all time-since-cut groups (mean difference flat-upright = -0.00602, SE = 0.0011, $p < 0.0001$). Upright frogs lost significantly more water than flat frogs during the day across all time-since-cut groups (flat – upright mean difference = -0.00548 % weight lost, SE = 0.0692, $p < .0001$). Differences in temperature across postures were not significantly different (flat – upright mean

difference = 0.00216 °C, SE = 0.0262, $p = 0.934$). Posture had no significant interactions with plot type, age, or microhabitat for any of the three response variables during the day.

At night, upright frogs had significantly higher performance (flat – upright mean difference = -0.00291, SE = 0.000764, $p < .0001$) and water loss (flat – upright mean difference = -0.28 % weight lost, SE = 0.0402, $p < .0001$) than flat frogs across all ages and plot types and water loss. There were no significant differences in average temperature (flat – upright mean difference = 0.000848 °C, SE = 0.0252, $p = 0.973$). Posture had no significant interactions with plot type, age, or microhabitat for any of the three response variables at night.

4 DISCUSSION

4.1 Diminished Performance in Clearcuts Overnight

Clearcuts provide warmer daytime conditions that align with higher gray treefrog performance, but this is offset by reduced structural complexity that accelerates evaporative water loss. However, at night when gray treefrogs are most active, performance is reduced in clearcut microhabitats, while forests maintain more stable overnight environments that support higher physiological performance, a result that is consistent with my second hypothesis. This reflects the colder nighttime temperatures in clearcuts, which approach lower thermal limits (Figure 3.7:C), imposing cold stress that reduces physiological function (M. E. Feder & Hofmann, 1999; Schmidt et al., 2024).

At night, despite greater overstory cover, gray treefrogs in forests actually experience higher water loss than in clearcuts. This increased dehydration likely results from the interaction of temperature, vapor pressure deficit (VPD), and vegetation structure where my replicas were placed (~ 2 m from the ground). Forests retain more heat at night but often have lower relative humidity under the canopy—especially in older-growth stands with large gaps and heightened

1116 exposure to drying winds (De Frenne et al., 2013). Increased airflow in these forests may disrupt
1117 boundary layers around frogs and enhance evaporative water loss. In contrast, clearcuts
1118 experience radiative cooling, lowering air temperatures and reducing the atmosphere's moisture-
1119 holding capacity (Chen et al., 1993). This often brings relative humidity closer to dew point and
1120 slows evaporation (De Frenne et al., 2013). Mid-successional clearcuts (10–20 years post-
1121 harvest) were observed to have dense shrub and deciduous regrowth, which restricts vertical air
1122 movement and traps humid, still air. These conditions promote thicker boundary layers and
1123 reduce evaporation, offering some buffer against dehydration despite being more exposed than
1124 closed-canopy forests (Peterman et al., 2013; Peterman & Semlitsch, 2014; Rittenhouse et al.,
1125 2008).

1126 Although nighttime water loss was greater in forests than clearcuts, performance was
1127 estimated to be higher in forests because overall water loss rates were too low to negatively
1128 influence performance in either forests or clearcuts. Since water loss rates do not reach critical
1129 thresholds overnight, performance is highly dependent on temperature in nighttime microhabitats
1130 and the small hydric benefit in clearcuts is offset by lower nighttime temperatures. At colder
1131 temperatures, performance may be reduced regardless of hydration state (Figure 3.4; (H. John-
1132 Alder et al., 1988; H. B. John-Alder et al., 1989; Navas, 1996). This suggests that although
1133 clearcut microhabitats may reduce overnight water loss, the benefit is negligible because frogs
1134 are already performance-limited by cold. Impaired performance in clearcuts at night will be
1135 enhanced because nighttime (and daytime) performance was lower in uncovered microhabitats
1136 and in upright (rather than water conservation) postures. Nocturnal gray treefrogs face a decision
1137 during peak activity hours between remaining inactive under thermally favourable cover to
1138 reduce physiological stress and emerging into colder, exposed microhabitats to forage and

replenish energy reserves. While amphibians often perform under various hydration and temperature states that do not maximize performance (Anderson & Andrade, 2017; Martin & Huey, 2008; A. Mitchell & Bergmann, 2016; Payne et al., 2015), if environmental temperatures consistently deviate from species' thermal optima, performance capacities such as locomotion, foraging efficiency, predator escape, and reproductive output can decline (Huey et al., 2009; Navas et al., 2016). My results indicate that this decision comes with greater costs in clearcut habitats, where reduced performance will make key activities, like foraging, more costly than in forested environments where overall performance is higher.

4.2 Beneficially heightened daytime temperatures render frogs more vulnerable to desiccation

Water loss, temperature, and performance were all higher in clearcuts than in forests during the day (Table 3.4, Table 3.5, Table 3.6). These results are consistent with previous research demonstrating elevated daytime temperatures and increased evaporative water loss in open-canopy habitats that receive more direct solar radiation (De Frenne et al., 2019; Perez-Navarro et al., 2024; Richard et al., 2021). In contrast to the lower performance observed overnight in clearcuts driven by thermal stress under colder conditions, gray treefrogs may benefit from increased daytime temperatures over relatively short periods of time (4 hours). These performance benefits may be of little ecological relevance since gray treefrogs are predominantly nocturnal and thus most likely to be inactive, in sheltered refuges, during the day. However, even in sheltered refuges, and while adopting water-conservation posture (Anderson & Andrade, 2017; Tracy et al., 2010), inactive frogs will experience greater water loss in clearcuts and thus may need to leave these refuges to rehydrate, incurring costs through increased energy expenditure and increased predation risk that are not present in forested refugia (Rittenhouse et

al., 2009). Moreover, exposure of inactive frogs to higher temperatures in clearcuts while sheltering can elevate metabolic rates and thus induce greater energetic expenditure during inactivity compared to cooler forest refugia, necessitating greater energy acquisition at night (Rollins-Smith & Le Sage, 2023).

It is possible that treefrogs could shift their activity time to take advantage of higher performance in daytime clearcuts, analogous to how some species have altered activity times in response to climate change (Doody et al., 2019; Sinervo et al., 2010). However, the benefits of such a shift may be lower than my results suggest because the full physiological consequences of daytime conditions may not be captured within my four-hour sampling period. The advantage of higher performance depends on dehydration levels not exceeding 15-20%, after which performance declines (Figure 3.4); these thresholds are consistent with those found for other species (Anderson & Andrade, 2017; Greenberg & Palen, 2021). This threshold is likely to be surpassed under continued exposure to elevated clearcut temperatures in high latitude areas with long summer daylight periods, especially because treefrogs may actively increase evaporative water loss to mitigate thermal stress through evaporative cooling when body temperatures rise (Sunday et al., 2014; Tracy et al., 2010, 2014) elevating dehydration risk even in shaded or vegetated microhabitats (Spotila et al., 1992; Tracy, 1976). In the most recent clearcuts (<5 years old), water loss levels reached 20–30% during the four-hour daytime window, resulting in the lowest observed performances in clearcuts across all time-since-cut groups. Therefore, unless gray treefrogs can access sufficient water to rehydrate throughout the day, performance will eventually decline in cuts due to excessive dehydration, rendering them less physiological suitable than forests.

Gray treefrogs may be able to behaviourally buffer negative performance effects of clearcuts by seeking out favourable refugia for hydrothermal regulation. At night, this would mean seeking out covered microhabitats for their warmth, while during the day it would rely on finding water sources for rehydration. At night, covered microhabitats in clearcuts were consistently warmer than uncovered microhabitats across all successional stages (Table 3.8) and therefore provide buffering from decreased temperature stress in exposed clearcuts. This means treefrogs face a decision during peak activity hours between remaining inactive under thermally stable cover to reduce physiological stress or emerging into colder, exposed microhabitats to forage and replenish energy reserves from heightened metabolic activity during the day. A lack of humid, daytime, refugia will render frogs more vulnerable to desiccation if temperatures and water loss levels exceed or near individual limits regardless of their activity levels (Cline & Hunter, 2014; Roznik et al., 2018). The abundance and configuration, i.e. accessibility, of refugia substantially influence the effectiveness, and energetic costs of thermoregulation (Sears et al., 2016; Sears & Angilletta, 2015). While there are no existing data on the availability of hydrothermal refugia in clearcuts, it is well established that they are hotter and drier during the day (Chen et al., 1993; Hocking & Semlitsch, 2008), suggesting that such refugia are likely to be few and far between, a situation that could be heightened by increased periods of drought under ongoing climate change (Soltani et al., 2024; Zhang et al., 2021). Even within clearcuts from 1-14-years-old, refugia may not diminish the thermal stresses of elevated daytime temperatures (Table 3.4). Therefore, frogs may still need to thermoregulate due to temperatures increases within clearcut microhabitats (Scheffers et al., 2013) through evaporative cooling and could be at risk for desiccation despite being covered and inactive.

4.3 Long-term persistence of clearcutting effects

The effects of harvesting on gray treefrog temperature, water loss and performance were still evident after 20 years post-harvest (Figure 3.7; Figure 3.6). The trajectory of post-harvest succession diverges markedly from that of undisturbed forest stands, not only in tree species composition but also in vertical structure and microclimatic buffering (Chen et al., 1993). These differences are particularly relevant given that structural recovery of microhabitats critical for many amphibian species may take over 20 years to establish (Figure 3.7). While some vegetative cover returns rapidly, the transition toward structural and compositional maturity in boreal forests can take between 50-60 years post-clearcut, and full ecological recovery may exceed 60 years depending on disturbance intensity and landscape context (Brassard & and Chen, 2006; Cyr et al., 2009). Given treefrogs relatively short (5-7 year) lifespans compared to microhabitat recovery periods (20+ years), clearcutting could threaten persistence of forest-dwelling biphasic frog populations by causing long-term reductions in the quantity and quality of suitable microhabitat conditions and by reducing landscape connectivity and permeability (Becker et al., 2007; Harper et al., 2008; Semlitsch, 2000) over multiple generations. Long-term effects of clearcutting could lead to reduced amphibian richness and abundance by increasing mortality, promoting evacuation of clearcut habitats, and declines in abundance (deMaynadier & Hunter Jr., 1995; Harper et al., 2015; Semlitsch et al., 2009; Todd et al., 2009; Todd & Rothermel, 2006). Importantly, juvenile dispersal is significantly hindered in early successional or structurally simplified habitats (Patrick et al., 2006). Without immediate recovery of microhabitat spaces, or following years of consecutive dry periods and warmer temperatures, recruitment could be suppressed over multiple generations and shorter-lived species may have no opportunities to recover (Harper et al., 2015).

1229 Additionally, altered habitat quality may lead to overcrowding in smaller forest patches if
1230 individuals were unwilling to occupy cleared habitats. This could intensify competition for food
1231 and shelter, increase stress and predation risk, and limit energy intake (Veysey Powell & Babbitt,
1232 2015). Energetic constraints could reduce investment in growth, reproduction, and storage,
1233 resulting in poorer body condition and smaller adult size across generations (Janin et al., 2011;
1234 Veysey Powell & Babbitt, 2015). Collectively, such energetic and demographic constraints can
1235 compound population declines in harvested landscapes

1236 The retention of negative performance impacts through succession may also impact
1237 metapopulation dynamics by making it more costly to move through the landscape to reach
1238 favourable forest patches or edges. Moving through areas where locomotory performance is
1239 lower will have greater energetic costs (Harper et al., 2015; Patrick et al., 2006; Rittenhouse et
1240 al., 2009; Todd et al., 2009), making dispersal to forest patches more challenging. These effects
1241 will be compounded by greater exposure to predation through increased movement frequency,
1242 and frogs' own reluctance to cross, more open habitats (Joly et al., 2003; Popescu & Hunter,
1243 2011; Rittenhouse et al., 2009). There is also the potential for synergistic effects where not only
1244 may dispersing frogs be more detectable to predators in early successional habitats, but will also
1245 have reduced escape ability due to impaired locomotory performance. This increase in landscape
1246 resistance may also operate on more micro-scales, increasing the cost of seeking out
1247 hydrothermal refugia in successional forest. Together, these constraints may hinder both long-
1248 distance dispersal between forest patches and fine-scale movements toward microhabitats
1249 offering thermal and hydric refuge, ultimately reducing recolonization rates, increasing local
1250 extirpation risk, and weakening overall metapopulation connectivity.

Current forest management guidelines in Ontario prioritize connectivity and habitat retention primarily for large mammals like moose and caribou (OMNRF, 2019), often overlooking the fine-scale requirements of smaller taxa with more constrained population ranges. While the OMNRF reports differences of opinion on the importance of landscape pattern versus habitat amount for biodiversity conservation, small-bodied or low-mobility species (like amphibians) are under-represented in the studies cited by OMNRF, even though they may respond differently to landscape pattern due to limited dispersal. Given the vulnerability of even disturbance-resistance amphibians like the gray treefrog to dehydration and thermal extremes (Demaynadier & Hunter, 1998; Harper et al., 2005), maintaining functional corridors of shaded, humid habitat between remnant patches and intact forests is essential for facilitating movement, recolonization, and genetic exchange (see Kuuluvainen, 2009; Kuuluvainen et al., 2021). Therefore, sustainable harvesting practices should consider not only the timing and intensity of silvicultural interventions but also the spatial configuration of harvested and retained areas. Incorporating connectivity planning for a broader range of taxa, especially small vertebrates and invertebrates, will improve the ecological integrity and resilience of regenerating forest landscapes.

4.4 Limitations and assumptions

My results come with several caveats. Firstly, I estimated the performance of gray treefrogs within microhabitats based on measurements of performance in the lab. However, these measures may not accurately capture hydrothermal performance relationships in the field. Additionally conditions that maximize performance may not be preferred conditions in the field (Martin & Huey, 2008; A. Mitchell & Bergmann, 2016). My results demonstrate the *capacity* for performance in relevant microhabitats frogs may encounter in the Boreal Forest transition zone.

However, frogs are subject to various other physiological and ecological considerations that may influence their behaviour and habitat use (Navas et al., 2021). They therefore need to make decisions about conserving energy versus expending energy for activities like foraging, predator avoidance, and hydrothermal-regulation. Frogs in drier and warmer habitats may sacrifice peak locomotor performance (needed for escaping predators or foraging) to reduce dehydration risk (A. Mitchell & Bergmann, 2016). There remains a critical gap in empirical assessments of how and when species prioritize one regulatory mechanism over the other, especially under seasonal or anthropogenically altered conditions where hydrothermal environments are shifting in opposing directions (Gunderson & Stillman, 2015; Navas et al., 2008, 2021). Further, I could not feasibly assess whether acclimation may alter hydrothermal performance curves (Schulte et al., 2011) and thus alter estimates of performance capacity across macro and microhabitats.

My sampling was also restricted in two important ways. First, it largely captures conditions in the mid-story. Most previous research on temperature and water loss has focused on ground-level microhabitats for terrestrial amphibians (Keppel et al., 2017), but for treefrogs conditions in the canopy are likely to be important (Olson et al., 2023). To capture all conditions that gray treefrogs may be facing, future work should explore microclimatic conditions in the mid to upper canopy, where treefrogs may exploit warmer air layers (Johnson et al., 2008; Laughlin et al., 2017; Olson et al., 2023). Measurements taken at a single level likely underestimate the buffering effects of mid-successional vegetation, especially above two meters where boundary layer dynamics and humidity conditions may differ from conditions in the mid to upper canopy layers. It also ignores the potential for vertical movement to allow for hydrothermal buffering (Klinges et al., 2024; Scheffers et al., 2013). Secondly, my sample was entirely male (because of practical limitations on finding females in the field), and past research

indicates that male and female performance may differ under similar physiological conditions due males' smaller body-size, which has been shown to influence metabolic rates, thermal inertia, and performance due to differences in optimal body temperatures (Kingsolver & Huey, 2008; Pottier et al., 2021; Rohr et al., 2018). This underscores the importance of understanding sex specific physiological and behavioral adaptations to predict responses to habitat alterations.

Finally, while my findings show that treefrogs have reduced performance in clearcuts and that these effects are sustained across multiple decades of succession, I do not have evidence that these performance impairments have implications at broader ecological scales, specifically population dynamics. Few studies of thermal vulnerability make this link (Gunderson & Stillman, 2015), but Sinervo et al. 2010 showed that a model in which degradation of thermal environments resulted in individuals behaviorally minimizing exposure to high daytime temperatures by reducing activity, could predict population declines and persistence. The limited research available on physiological costs in clearcut environments have identified reduced amphibian survival and body conditions (Mazerolle et al., 2021; Rittenhouse et al., 2009; Todd et al., 2014; Veysey Powell & Babbitt, 2015), suggesting that the performance costs associated with clearcutting we observed for gray treefrogs in the boreal-transition zone could plausibly influence population abundance and persistence.

5 CONCLUSION

While clearcuts may offer short-term thermal advantages during the day that align with the thermal optima of gray treefrogs, these benefits are counteracted by elevated evaporative water loss due to increased exposure and lack of hydrothermally preferable refugia. At night, when gray treefrogs are most active, performance capacity is consistently lower in clearcut environments, with forests providing more stable microclimates that better support physiological

function. These microclimatic effects persist even two decades post-harvest, underscoring the long-term physiological costs of clearcutting in forest ecosystems. While climate change alters average conditions over large spatial and temporal scales, habitat modification can produce abrupt and localized microclimatic extremes that directly impact physiological function. These structural changes often go unrecognized in assessments of amphibian vulnerability, despite their potent effects on individual performance and behavioral flexibility. My results highlight the need to incorporate the joint effects of temperature and hydration on microhabitat quality and physiological performance into conservation assessments. Future work should focus on long-term, multi-generational studies of amphibian population recovery in repeatedly disturbed or slowly regenerating forests. While some research has examined amphibian presence across forest age classes (e.g., Patrick et al., 2006; Harper et al., 2015), most have examined abundance immediately post-disturbance, but few have followed population recovery trajectories over time through forest recovery, limiting our understanding of how chronic habitat degradation influences long-term population persistence and dispersal.

6 REFERENCES

- Anderson, R. C. O., & Andrade, D. V. (2017). Trading heat and hops for water: Dehydration effects on locomotor performance, thermal limits, and thermoregulatory behavior of a terrestrial toad. *Ecology and Evolution*, 7(21), 9066–9075. <https://doi.org/10.1002/ece3.3219>
- Angilletta, M. J. (2006). Estimating and comparing thermal performance curves. *Journal of Thermal Biology*, 31(7), 541–545. <https://doi.org/10.1016/j.jtherbio.2006.06.002>
- Angilletta, M. J. (2009). *Thermal adaptation: A theoretical and empirical synthesis*. Oxford University Press.
- Anoszko, E., Frelich, L. E., Rich, R. L., & Reich, P. B. (2022). Wind and fire: Rapid shifts in tree community composition following multiple disturbances in the southern boreal forest. *Ecosphere*, 13(3), e3952. <https://doi.org/10.1002/ecs2.3952>
- Anyomi, K. A., Neary, B., Chen, J., & Mayor, S. J. (2022). A critical review of successional dynamics in boreal forests of North America. *Environmental Reviews*, 30(4), 563–594. <https://doi.org/10.1139/er-2021-0106>
- Ashrafi, R., Bruneaux, M., Sundberg, L., Pulkkinen, K., Valkonen, J., & Ketola, T. (2018). Broad thermal tolerance is negatively correlated with virulence in an opportunistic bacterial pathogen. *Evolutionary Applications*, 11(9), 1700–1714. <https://doi.org/10.1111/eva.12673>
- Becker, C. G., Fonseca, C. R., Haddad, C. F. B., Batista, R. F., & Prado, P. I. (2007). Habitat Split and the Global Decline of Amphibians. *Science*, 318(5857), 1775–1777. <https://doi.org/10.1126/science.1149374>

1365 Bennett, A. F. (1978). Activity Metabolism of the Lower Vertebrates. *Annual Review of*
 1366 *Physiology*, 40(Volume 40, 1978), 447–469.
 1367 <https://doi.org/10.1146/annurev.ph.40.030178.002311>

1368 Bergeron, Y., & Fenton, N. J. (2012). Boreal forests of eastern Canada revisited: Old growth,
 1369 nonfire disturbances, forest succession, and biodiversity. *Botany*, 90(6), 509–523.
 1370 <https://doi.org/10.1139/b2012-034>

1371 Bergeron, Y., Harvey, B., Leduc, A., & Gauthier, S. (1999). Forest management guidelines
 1372 based on natural disturbance dynamics: Stand- and forest-level considerations. *The*
 1373 *Forestry Chronicle*, 75(1), 49–54. <https://doi.org/10.5558/tfc75049-1>

1374 Biazzo, I., & Quintana-Ascencio, P. (2022). Canopies, the Final Frog-tier: Exploring responses
 1375 of a specialist treefrog to prescribed fire in a pyrogenic ecosystem. *FIRE ECOLOGY*,
 1376 18(1). <https://doi.org/10.1186/s42408-022-00148-1>

1377 Boucher, Y., St-Laurent, M.-H., & Grondin, P. (2011). Logging-Induced Edge and Configuration
 1378 of Old-Growth Forest Remnants in the Eastern North American Boreal Forests. *Natural*
 1379 *Areas Journal*, 31(3), 300–306. <https://doi.org/10.3375/043.031.0313>

1380 Brandt, J. P., Flannigan, M. D., Maynard, D. G., Thompson, I. D., & Volney, W. J. A. (2013).
 1381 An introduction to Canada’s boreal zone: Ecosystem processes, health, sustainability, and
 1382 environmental issues. *Environmental Reviews*, 21(4), 207–226.
 1383 <https://doi.org/10.1139/er-2013-0040>

1384 Brassard, B. W., & and Chen, H. Y. H. (2006). Stand Structural Dynamics of North American
 1385 Boreal Forests. *Critical Reviews in Plant Sciences*, 25(2), 115–137.
 1386 <https://doi.org/10.1080/07352680500348857>

1387 Briere, J.-F., Pracros, P., Le Roux, A.-Y., & Pierre, J.-S. (1999). A Novel Rate Model of
 1388 Temperature-Dependent Development for Arthropods. *Environmental Entomology*,
 1389 28(1), 22–29. <https://doi.org/10.1093/ee/28.1.22>

1390 Brooks, M. E., Kristensen, K., Benthem, K. V. J., Magnusson, A., Berg, C. W., Nielsen, A.,
 1391 Skaug, H. J., Maechler, M., & Bolker, B. M. (2017). glmmTMB Balances Speed and
 1392 Flexibility Among Packages for Zero-inflated Generalized Linear Mixed Modeling. *The*
 1393 *R Journal*, 9(2), 378–400. <https://doi.org/10.32614/RJ-2017-066>.

1394 Burnham, K. P., & Anderson, D. R. (Eds.). (2004). *Model Selection and Multimodel Inference*.
 1395 Springer. <https://doi.org/10.1007/b97636>

1396 Burrow, A. K., McEntire, K. D., & Maerz, J. C. (2023). Estimating the potential drivers of
 1397 dispersal outcomes for juvenile gopher frogs (*Rana capito*) using agent-based models.
 1398 *Frontiers in Ecology and Evolution*, 11, 1026541.
 1399 <https://doi.org/10.3389/fevo.2023.1026541>

1400 Campbell, G. S., & Norman, J. M. (1998). *An introduction to environmental biophysics*. Springer
 1401 Science and Business Media Inc.

1402 Carleton, T. J., & Maclellan, P. (1994). Woody vegetation responses to fire versus clear-cutting
 1403 logging: A comparative survey in the central Canadian boreal forest. *Écoscience*, 1(2),
 1404 141–152. <https://doi.org/10.1080/11956860.1994.11682238>

1405 Chen, J. (1999). Microclimate in Forest Ecosystem and Landscape Ecology. *BioScience*, 49,
 1406 288–297.

1407 Chen, J., Franklin, J. F., & Spies, T. A. (1993). Contrasting microclimates among clearcut, edge,
 1408 and interior of old-growth Douglas-fir forest. *Agricultural and Forest Meteorology*, 63,
 1409 219–237.

1410 Cline, B. B., & Hunter, M. L. (2014). Different open-canopy vegetation types affect matrix
 1411 permeability for a dispersing forest amphibian. *Journal of Applied Ecology*, 51(2), 319–
 1412 329. <https://doi.org/10.1111/1365-2664.12197>

1413 Cossins, A. (2012). *Temperature Biology of Animals*. Springer Science & Business Media.

1414 Crins, W. J., Gray, P. A., Uhlig, P. W. C., & Wester, M. C. (2024). *The Ecosystems of Ontario,*
 1415 *Part 1: Ecozones and Ecoregions*.

1416 Cyr, D., Gauthier, S., Bergeron, Y., & Carcaillet, C. (2009). Forest management is driving the
 1417 eastern North American boreal forest outside its natural range of variability. *Frontiers in*
 1418 *Ecology and the Environment*, 7(10), 519–524. <https://doi.org/10.1890/080088>

1419 Damos, P. T., & Savopoulou-Soultani, M. (2008). Temperature-dependent bionomics and
 1420 modeling of *Anarsia lineatella* (Lepidoptera: Gelechiidae) in the laboratory. *Journal of*
 1421 *Economic Entomology*, 101(5), 1557–1567. [https://doi.org/10.1603/0022-](https://doi.org/10.1603/0022-0493(2008)101[1557:tbamoa]2.0.co;2)
 1422 [0493\(2008\)101\[1557:tbamoa\]2.0.co;2](https://doi.org/10.1603/0022-0493(2008)101[1557:tbamoa]2.0.co;2)

1423 Dawe, D. A., Parisien, M.-A., Van Dongen, A., & Whitman, E. (2022). Initial succession after
 1424 wildfire in dry boreal forests of northwestern North America. *Plant Ecology*, 223(7),
 1425 789–809. <https://doi.org/10.1007/s11258-022-01237-6>

1426 De Frenne, P., Rodríguez-Sánchez, F., Coomes, D. A., Baeten, L., Verstraeten, G., Vellend, M.,
 1427 Bernhardt-Römermann, M., Brown, C. D., Brunet, J., Cornelis, J., Decocq, G. M.,
 1428 Dierschke, H., Eriksson, O., Gilliam, F. S., Hédli, R., Heinken, T., Hermy, M., Hommel,
 1429 P., Jenkins, M. A., ... Verheyen, K. (2013). Microclimate moderates plant responses to
 1430 macroclimate warming. *Proceedings of the National Academy of Sciences of the United*
 1431 *States of America*, 110(46), 18561–18565. <https://doi.org/10.1073/pnas.1311190110>

1432 De Frenne, P., Zellweger, F., Rodríguez-Sánchez, F., Scheffers, B. R., Hylander, K., Luoto, M.,
 1433 Vellend, M., Verheyen, K., & Lenoir, J. (2019). Global buffering of temperatures under
 1434 forest canopies. *Nature Ecology & Evolution*, 3(5), 744–749.
 1435 <https://doi.org/10.1038/s41559-019-0842-1>
 1436 deMaynadier, P. G., & Hunter Jr., M. L. (1995). The relationship between forest management
 1437 and amphibian ecology: A review of the North American literature. *Environmental*
 1438 *Reviews*, 3(3–4), 230–261. <https://doi.org/10.1139/a95-012>
 1439 Demaynadier, P. G., & Hunter, M. L. (1998). Effects of Silvicultural Edges on the Distribution
 1440 and Abundance of Amphibians in Maine. *Conservation Biology*, 12(2), 340–352.
 1441 <https://doi.org/10.1111/j.1523-1739.1998.96412.x>
 1442 Deutsch, C. A. (2008). Impacts of climate warming on terrestrial ectotherms across latitude.
 1443 *Proceedings of the National Academy of Sciences, U.S.A*, 105, 6668–6672.
 1444 Dobrowski, S. Z. (2011). A climatic basis for microrefugia: The influence of terrain on climate.
 1445 *Global Change Biology*, 17(2), 1022–1035. [https://doi.org/10.1111/j.1365-](https://doi.org/10.1111/j.1365-2486.2010.02263.x)
 1446 [2486.2010.02263.x](https://doi.org/10.1111/j.1365-2486.2010.02263.x)
 1447 Dodd, C. K. (2013). *Frogs of the United States and Canada*.
 1448 Doody, J. S., McHenry, C. R., Rhind, D., & Clulow, S. (2019). Novel habitat causes a shift to
 1449 diurnal activity in a nocturnal species. *Scientific Reports*, 9(1), 230.
 1450 <https://doi.org/10.1038/s41598-018-36384-2>
 1451 Dormann, C. F., Calabrese, J. M., Guillera-Aroita, G., Matechou, E., Bahn, V., Bartoń, K.,
 1452 Beale, C. M., Ciuti, S., Elith, J., Gerstner, K., Guelat, J., Keil, P., Lahoz-Monfort, J. J.,
 1453 Pollock, L. J., Reineking, B., Roberts, D. R., Schröder, B., Thuiller, W., Warton, D. I., ...
 1454 Hartig, F. (2018). Model averaging in ecology: A review of Bayesian, information-

1455 theoretic, and tactical approaches for predictive inference. *Ecological Monographs*,
 1456 88(4), 485–504. <https://doi.org/10.1002/ecm.1309>
 1457 Duarte, H., Tejedo, M., Katzenberger, M., Marangoni, F., Baldo, D., Beltrán, J. F., Martí, D. A.,
 1458 Richter-Boix, A., & Gonzalez-Voyer, A. (2012). Can amphibians take the heat?
 1459 Vulnerability to climate warming in subtropical and temperate larval amphibian
 1460 communities. *Global Change Biology*, 18(2), 412–421. <https://doi.org/10.1111/j.1365->
 1461 2486.2011.02518.x
 1462 Dupuch, A., & Fortin, D. (2013). The extent of edge effects increases during post-harvesting
 1463 forest succession. *Biological Conservation*, 162, 9–16.
 1464 <https://doi.org/10.1016/j.biocon.2013.03.023>
 1465 Economou, L., & Kontodimas, D. (2004). Comparative Temperature-Dependent Development of
 1466 *Nephus includens* (Kirsch) and *Nephus bisignatus* (Boheman) (Coleoptera:
 1467 Coccinellidae) Preying on *Planococcus citri* (Risso) (Homoptera: Pseudococcidae):
 1468 Evaluation of a Linear and Various Nonlinear Models Using Specific Criteria.
 1469 *Environmental Entomology*.
 1470 https://www.academia.edu/20568703/Comparative_Temperature_Dependent_Development_of_Nephus_includens_Kirsch_and_Nephus_bisignatus_Boheman_Coleoptera_Coccinellidae_Preying_on_Planococcus_citri_Risso_Homoptera_Pseudococcidae_Evaluation_of_a_Linear_and_Various_Nonlinear_Models_Using_Specific_Criteria
 1471
 1472
 1473
 1474 Elith, J., & Leathwick, J. R. (2009). Species Distribution Models: Ecological Explanation and
 1475 Prediction Across Space and Time. *Annual Review of Ecology, Evolution, and*
 1476 *Systematics*, 40(Volume 40, 2009), 677–697.
 1477 <https://doi.org/10.1146/annurev.ecolsys.110308.120159>

1478 Eubank, W. P., Atmar, J. W., & Ellington, J. J. (1973). The Significance and Thermodynamics of
 1479 Fluctuating Versus Static Thermal Environments on *Heliothis zea* Egg Development
 1480 Rates 1. *Environmental Entomology*, 2(4), 491–496. <https://doi.org/10.1093/ee/2.4.491>
 1481 Feder, M. E., & Hofmann, G. E. (1999). HEAT-SHOCK PROTEINS, MOLECULAR
 1482 CHAPERONES, AND THE STRESS RESPONSE: Evolutionary and Ecological
 1483 Physiology. *Annual Review of Physiology*, 61(Volume 61, 1999), 243–282.
 1484 <https://doi.org/10.1146/annurev.physiol.61.1.243>
 1485 Fenton, N. J., Simard, M., & Bergeron, Y. (2009). Emulating natural disturbances: The role of
 1486 silviculture in creating even-aged and complex structures in the black spruce boreal forest
 1487 of eastern North America. *Journal of Forest Research*, 14(5), 258–267.
 1488 <https://doi.org/10.1007/s10310-009-0134-8>
 1489 Franklin, J. F., Spies, T. A., Pelt, R. V., Carey, A. B., Thornburgh, D. A., Berg, D. R.,
 1490 Lindenmayer, D. B., Harmon, M. E., Keeton, W. S., Shaw, D. C., Bible, K., & Chen, J.
 1491 (2002). Disturbances and structural development of natural forest ecosystems with
 1492 silvicultural implications, using Douglas-fir forests as an example. *Forest Ecology and*
 1493 *Management*, 155(1), 399–423. [https://doi.org/10.1016/S0378-1127\(01\)00575-8](https://doi.org/10.1016/S0378-1127(01)00575-8)
 1494 Frazier, M. R., Huey, R. B., & Berrigan, D. (2006). Thermodynamics constrains the evolution of
 1495 insect population growth rates: “warmer is better.” *The American Naturalist*, 168(4),
 1496 512–520. <https://doi.org/10.1086/506977>
 1497 Gardner, T. A., Barlow, J., & Peres, C. A. (2007). Paradox, presumption and pitfalls in
 1498 conservation biology: The importance of habitat change for amphibians and reptiles.
 1499 *Biological Conservation*, 138(1–2), 166–179.
 1500 <https://doi.org/10.1016/j.biocon.2007.04.017>

1501 Gatten, R. E. (1987). Activity Metabolism of Anuran Amphibians: Tolerance to Dehydration.
 1502 *Physiological Zoology*, 60(5), 576–585. <https://doi.org/10.1086/physzool.60.5.30156131>
 1503 Gatten, R. E., Miller, K., & Full, R. J. (1992). Energetics at Rest and during Locomotion. In M.
 1504 E. Bicudo & W. W. Burggren (Eds.), *Environmental Physiology of the Amphibians*. (pp.
 1505 314–377). University of Chicago Press.
 1506 Gebhardt, A., Bivand, R., & library), D. S. (author of the shull. (2024). *interp: Interpolation*
 1507 *Methods* (Version 1.1-6) [Computer software]. [https://cran.r-](https://cran.r-project.org/web/packages/interp/index.html)
 1508 [project.org/web/packages/interp/index.html](https://cran.r-project.org/web/packages/interp/index.html)
 1509 Geiger R., Aron R. H., & Todhunter P. (1965). *The Climate Near the Ground*. Harvard
 1510 University Press.
 1511 Grandpré, L., Gagnon, D., & Bergeron, Y. (1993). Changes in the understory of Canadian
 1512 southern boreal forest after fire. *Journal of Vegetation Science*, 4(6), 803–810.
 1513 <https://doi.org/10.2307/3235618>
 1514 Greenberg, D. A., & Palen, W. J. (2021). Hydrothermal physiology and climate vulnerability in
 1515 amphibians. *Proceedings of the Royal Society B: Biological Sciences*, 288(1945),
 1516 20202273. <https://doi.org/10.1098/rspb.2020.2273>
 1517 Gunderson, A. R., & Stillman, J. H. (2015). Plasticity in thermal tolerance has limited potential
 1518 to buffer ectotherms from global warming. *Proceedings of the Royal Society B:*
 1519 *Biological Sciences*, 282(1808), 20150401. <https://doi.org/10.1098/rspb.2015.0401>
 1520 Harper, E. B., Patrick, D. A., & Gibbs, J. P. (2015). Impact of forestry practices at a landscape
 1521 scale on the dynamics of amphibian populations. *Ecological Applications: A Publication*
 1522 *of the Ecological Society of America*, 25(8), 2271–2284. [https://doi.org/10.1890/14-](https://doi.org/10.1890/14-0962.1)
 1523 [0962.1](https://doi.org/10.1890/14-0962.1)

1524 Harper, E. B., Rittenhouse, T. a. G., & Semlitsch, R. D. (2008). Demographic Consequences of
 1525 Terrestrial Habitat Loss for Pool-Breeding Amphibians: Predicting Extinction Risks
 1526 Associated with Inadequate Size of Buffer Zones. *Conservation Biology*, 22(5), 1205–
 1527 1215. <https://doi.org/10.1111/j.1523-1739.2008.01015.x>
 1528 Harper, K. A., Macdonald, S. E., Burton, P. J., Chen, J., Brosfokske, K. D., Saunders, S. C.,
 1529 Euskirchen, E. S., Roberts, D., Jaiteh, M. S., & Esseen, P. (2005). Edge Influence on
 1530 Forest Structure and Composition in Fragmented Landscapes. *Conservation Biology*,
 1531 19(3), 768–782. <https://doi.org/10.1111/j.1523-1739.2005.00045.x>
 1532 Hartig, F., Lohse, L., & leite, M. de S. (2024). *DHARMa: Residual Diagnostics for Hierarchical*
 1533 *(Multi-Level / Mixed) Regression Models* (Version 0.4.7) [Computer software].
 1534 <https://cran.r-project.org/web/packages/DHARMa/index.html>
 1535 Hastings, T. P., Hossack, B. R., Fishback, L., & Davenport, J. M. (2023). Using physiological
 1536 conditions to assess current and future habitat use of a Subarctic frog. *Integrative*
 1537 *Zoology*, 18(1), 2–14. <https://doi.org/10.1111/1749-4877.12649>
 1538 Hermosilla, T., Wulder, M. A., White, J. C., Coops, N. C., Hobart, G. W., & Campbell, L. B.
 1539 (2016). Mass data processing of time series Landsat imagery: Pixels to data products for
 1540 forest monitoring. *International Journal of Digital Earth*, 9(11), 1035–1054.
 1541 <https://doi.org/10.1080/17538947.2016.1187673>
 1542 Hillman, S. S. (1978). The roles of oxygen delivery and electrolyte levels in the dehydrational
 1543 death of *Xenopus laevis*. *Journal of Comparative Physiology ? B*, 128(2), 169–175.
 1544 <https://doi.org/10.1007/BF00689481>

- 1545 Hillman, S. S., Withers, P. C., Drewes, R. C., & Hillyard, S. D. (2008). *Ecological and*
 1546 *Environmental Physiology of Amphibians*. Oxford University Press.
 1547 <https://doi.org/10.1093/acprof:oso/9780198570325.001.0001>
- 1548 Hocking, D. J., & Semlitsch, R. D. (2008). Effects of Experimental Clearcut Logging on Gray
 1549 Treefrog (*Hyla versicolor*) Tadpole Performance. *Journal of Herpetology*, 42(4), 689.
 1550 <https://doi.org/10.1670/07-294R1.1>
- 1551 Hof, C., Araújo, M. B., Jetz, W., & Rahbek, C. (2011). Additive threats from pathogens, climate
 1552 and land-use change for global amphibian diversity. *Nature*, 480(7378), 516–519.
 1553 <https://doi.org/10.1038/nature10650>
- 1554 Hossack, B. R., Eby, L. A., Guscio, C. G., & Corn, P. S. (2009). Thermal characteristics of
 1555 amphibian microhabitats in a fire-disturbed landscape. *Forest Ecology and Management*,
 1556 258(7), 1414–1421. <https://doi.org/10.1016/j.foreco.2009.06.043>
- 1557 Hossack, B. R., Lowe, W. H., Honeycutt, R. K., Parks, S. A., & Corn, P. S. (2013). Interactive
 1558 effects of wildfire, forest management, and isolation on amphibian and parasite
 1559 abundance. *Ecological Applications*, 23(2), 479–492. <https://doi.org/10.1890/12-0316.1>
- 1560 Huey, R. B., Deutsch, C. A., Tewksbury, J. J., Vitt, L. J., Hertz, P. E., Álvarez Pérez, H. J., &
 1561 Garland, T. (2009). Why tropical forest lizards are vulnerable to climate warming.
 1562 *Proceedings of the Royal Society B: Biological Sciences*, 276(1664), 1939–1948.
 1563 <https://doi.org/10.1098/rspb.2008.1957>
- 1564 Huey, R. B., Kearney, M. R., Krockenberger, A., Holtum, J. A. M., Jess, M., & Williams, S. E.
 1565 (2012). Predicting organismal vulnerability to climate warming: Roles of behaviour,
 1566 physiology and adaptation. *Philosophical Transactions of the Royal Society B: Biological*
 1567 *Sciences*, 367(1596), 1665–1679. <https://doi.org/10.1098/rstb.2012.0005>

1568 Huey, R. B., & Kingsolver, J. G. (1989). Evolution of thermal sensitivity of ectotherm
 1569 performance. *Trends in Ecology & Evolution*, 4(5), 131–135.
 1570 [https://doi.org/10.1016/0169-5347\(89\)90211-5](https://doi.org/10.1016/0169-5347(89)90211-5)

1571 Huey, R. B., & Stevenson, R. D. (1979). Integrating Thermal Physiology and Ecology of
 1572 Ectotherms: A Discussion of Approaches. *American Zoologist*, 19(1), 357–366.
 1573 <https://doi.org/10.1093/icb/19.1.357>

1574 Janin, A., Léna, J.-P., & Joly, P. (2011). Beyond occurrence: Body condition and stress hormone
 1575 as integrative indicators of habitat availability and fragmentation in the common toad.
 1576 *Biological Conservation*, 144(3), 1008–1016.
 1577 <https://doi.org/10.1016/j.biocon.2010.12.009>

1578 Janisch, E. (1925). Über die Temperaturabhängigkeit biologischer Vorgänge und ihre
 1579 kurvenmäßige Analyse. *Pflüger's Archiv für die gesamte Physiologie des Menschen und*
 1580 *der Tiere*, 209(1), 414–436. <https://doi.org/10.1007/BF01730929>

1581 Jõgiste, K., Korjus, H., Stanturf, J. A., Frelich, L. E., Baders, E., Donis, J., Jansons, A., Kangur,
 1582 A., Köster, K., Laarmann, D., Maaten, T., Marozas, V., Metslaid, M., Nigul, K.,
 1583 Polyachenko, O., Randveer, T., & Vodde, F. (2017). Hemiboreal forest: Natural
 1584 disturbances and the importance of ecosystem legacies to management. *Ecosphere*, 8(2),
 1585 e01706. <https://doi.org/10.1002/ecs2.1706>

1586 John-Alder, H. B., Barnhart, M. C., & Bennett, A. F. (1989). Thermal Sensitivity of Swimming
 1587 Performance and Muscle Contraction in Northern and Southern Populations of Tree
 1588 Frogs (*Hyla Crucifer*). *Journal of Experimental Biology*, 142(1), 357–372.
 1589 <https://doi.org/10.1242/jeb.142.1.357>

1590 John-Alder, H., Morin, P., & Lawler, S. (1988). Thermal Physiology, Phenology, and
 1591 Distribution of Tree Frogs. *American Naturalist - AMER NATURALIST*, 132.
 1592 <https://doi.org/10.1086/284868>
 1593 Johnson, J. B., & Omland, K. S. (2004). Model selection in ecology and evolution. *Trends in*
 1594 *Ecology & Evolution*, 19(2), 101–108. <https://doi.org/10.1016/j.tree.2003.10.013>
 1595 Johnson, J. R., Mahan, R. D., & Semlitsch, R. D. (2008). Seasonal Terrestrial Microhabitat Use
 1596 by Gray Treefrogs (*Hyla versicolor*) in Missouri Oak-hickory Forests. *Herpetologica*,
 1597 64(3), 259–269. <https://doi.org/10.1655/07-064.1>
 1598 Joly, P., Morand, C., & Cohas, A. (2003). Habitat fragmentation and amphibian conservation:
 1599 Building a tool for assessing landscape matrix connectivity. *Comptes Rendus Biologies*,
 1600 326, 132–139. [https://doi.org/10.1016/S1631-0691\(03\)00050-7](https://doi.org/10.1016/S1631-0691(03)00050-7)
 1601 Kamykowski, D. (1985). A survey of protozoan laboratory temperature studies applied to marine
 1602 dinoflagellate behavior from a field perspective. *Contributions in Marine Science*.
 1603 [https://www.researchgate.net/publication/252320931_A_survey_of_protozoan_laborator](https://www.researchgate.net/publication/252320931_A_survey_of_protozoan_laboratory_temperature_studies_applied_to_marine_dinoflagellate_behavior_from_the_viewpoint_of_ecological_implications_Appendix_Kamykowski_D_and_F_G_Giesbrecht_Confidence_limits_)
 1604 [y_temperature_studies_applied_to_marine_dinoflagellate_behavior_from_the_viewpoint](https://www.researchgate.net/publication/252320931_A_survey_of_protozoan_laboratory_temperature_studies_applied_to_marine_dinoflagellate_behavior_from_the_viewpoint_of_ecological_implications_Appendix_Kamykowski_D_and_F_G_Giesbrecht_Confidence_limits_)
 1605 [_of_ecological_implications_Appendix_Kamykowski_D_and_F_G_Giesbrecht_Confide](https://www.researchgate.net/publication/252320931_A_survey_of_protozoan_laboratory_temperature_studies_applied_to_marine_dinoflagellate_behavior_from_the_viewpoint_of_ecological_implications_Appendix_Kamykowski_D_and_F_G_Giesbrecht_Confidence_limits_)
 1606 [nce_limits_](https://www.researchgate.net/publication/252320931_A_survey_of_protozoan_laboratory_temperature_studies_applied_to_marine_dinoflagellate_behavior_from_the_viewpoint_of_ecological_implications_Appendix_Kamykowski_D_and_F_G_Giesbrecht_Confidence_limits_)
 1607 Kearney, M., & Porter, W. P. (2004). MAPPING THE FUNDAMENTAL NICHE:
 1608 PHYSIOLOGY, CLIMATE, AND THE DISTRIBUTION OF A NOCTURNAL
 1609 LIZARD. *Ecology*, 85(11), 3119–3131. <https://doi.org/10.1890/03-0820>
 1610 Kearney, M., & Predavec, M. (2000). DO NOCTURNAL ECTOTHERMS
 1611 THERMOREGULATE? A STUDY OF THE TEMPERATE GECKO *CHRISTINUS*

1612 *MARMORATUS. Ecology*, 81(11), 2984–2996. <https://doi.org/10.1890/0012->

1613 9658(2000)081[2984:DNETAS]2.0.CO;2

1614 Keenan, R. J., & Kimmins, J. P. (1993). The ecological effects of clear-cutting. *Environmental*

1615 *Reviews*, 1, 121–144.

1616 Keppel, G., Anderson, S., Williams, C., Kleindorfer, S., & O’Connell, C. (2017). Microhabitats

1617 and canopy cover moderate high summer temperatures in a fragmented Mediterranean

1618 landscape. *PLoS ONE*, 12(8), e0183106. <https://doi.org/10.1371/journal.pone.0183106>

1619 Kingsolver, J. G., & Huey, R. B. (2008). Size, temperature, and fitness: Three rules.

1620 *Evolutionary Ecology Research*, 10, 251–268.

1621 Klinges, D. H., Randriambololona, T., Lange, Z. K., Laterza-Barbosa, J., Randrianandrasana, H.,

1622 & Scheffers, B. R. (2024). Vertical and diel niches modulate thermal selection by

1623 rainforest frogs. *Proceedings of the Royal Society B: Biological Sciences*, 291(2034),

1624 20241497. <https://doi.org/10.1098/rspb.2024.1497>

1625 Köhler, A., Sadowska, J., Olszewska, J., Trzeciak, P., Berger-Tal, O., & Tracy, C. R. (2011).

1626 Staying warm or moist? Operative temperature and thermal preferences of common frogs

1627 (*Rana temporaria*). *Herpetological Journal*, 21, 17–26.

1628 Kontopoulos, D.-G., Sentis, A., Daufresne, M., Glazman, N., Dell, A. I., & Pawar, S. (2024). No

1629 universal mathematical model for thermal performance curves across traits and

1630 taxonomic groups. *Nature Communications*, 15(1), 8855. [https://doi.org/10.1038/s41467-](https://doi.org/10.1038/s41467-024-53046-2)

1631 024-53046-2

1632 Krenek, S., Berendonk, T. U., & Petzoldt, T. (2011). Thermal performance curves of

1633 *Paramecium caudatum*: A model selection approach. *European Journal of Protistology*,

1634 47(2), 124–137. <https://doi.org/10.1016/j.ejop.2010.12.001>

1635 Kuuluvainen, T. (2009). Forest Management and Biodiversity Conservation Based on Natural
 1636 Ecosystem Dynamics in Northern Europe: The Complexity Challenge. *AMBIO: A*
 1637 *Journal of the Human Environment*, 38(6), 309–315. <https://doi.org/10.1579/08-A-490.1>
 1638 Kuuluvainen, T., Angelstam, P., Frelich, L., Jõgiste, K., Koivula, M., Kubota, Y., Lafleur, B., &
 1639 Macdonald, E. (2021). Natural Disturbance-Based Forest Management: Moving Beyond
 1640 Retention and Continuous-Cover Forestry. *Frontiers in Forests and Global Change*, 4.
 1641 <https://doi.org/10.3389/ffgc.2021.629020>
 1642 Lactin, D. J., Holliday, N. J., Johnson, D. L., & Craigen, R. (1995). Improved rate model of
 1643 temperature-dependent development by arthropods.
 1644 *Environmental Entomology* 24: 68-75.
 1645 Larocque, G. R., Bell, F. W., Searle, E. B., Mayor, S. J., Schiks, T., & Kalantari, P. (2024).
 1646 Simulating the Long-Term Response of Forest Succession to Climate Change in the
 1647 Boreal Forest of Northern Ontario, Canada. *Forests*, 15(8), Article 8.
 1648 <https://doi.org/10.3390/f15081417>
 1649 Laughlin, M. M., Olson, E. R., & Martin, J. G. (2017). Arboreal camera trapping expands *Hyla*
 1650 *versicolor* complex (Hylidae) canopy use to new heights. *Ecology*, 98(8), 2221–2223.
 1651 <https://doi.org/10.1002/ecy.1843>
 1652 Lenth, R. V., Banfai, B., Bolker, B., Buerkner, P., Giné-Vázquez, I., Herve, M., Jung, M., Love,
 1653 J., Miguez, F., Piaskowski, J., Riebl, H., & Singmann, H. (2025). *emmeans: Estimated*
 1654 *Marginal Means, aka Least-Squares Means* (Version 1.11.2) [Computer software].
 1655 <https://cran.r-project.org/web/packages/emmeans/index.html>

1656 Lertzman-Lepofsky, G. F., Kissel, A. M., Sinervo, B., & Palen, W. J. (2020). Water loss and
 1657 temperature interact to compound amphibian vulnerability to climate change. *Glob.*
 1658 *Change Biol*, 26, 4868–4879.

1659 Lindenmayer, D. B., Kooyman, R. M., Taylor, C., Ward, M., & Watson, J. E. M. (2020). Recent
 1660 Australian wildfires made worse by logging and associated forest management. *Nature*
 1661 *Ecology & Evolution*, 4(7), 898–900. <https://doi.org/10.1038/s41559-020-1195-5>

1662 Logan, J. A., Wollkind, D. J., Hoyt, S. C., & Tanigoshi, L. K. (1976). An Analytic Model for
 1663 Description of Temperature Dependent Rate Phenomena in Arthropods 1. *Environmental*
 1664 *Entomology*, 5(6), 1133–1140. <https://doi.org/10.1093/ee/5.6.1133>

1665 Lüdecke, D., Ben-Shachar, M. S., Patil, I., Waggoner, P., & Makowski, D. (2021). performance:
 1666 An R Package for Assessment, Comparison and Testing of Statistical Models. *Journal of*
 1667 *Open Source Software*, 6(60), 3139. <https://doi.org/10.21105/joss.03139>

1668 Luedtke, J. A., Chanson, J., Neam, K., Hobin, L., Maciel, A. O., Catenazzi, A., Borzée, A.,
 1669 Hamidy, A., Aowphol, A., Jean, A., Sosa-Bartuano, Á., Fong G., A., de Silva, A.,
 1670 Fouquet, A., Angulo, A., Kidov, A. A., Muñoz Saravia, A., Diesmos, A. C., Tominaga,
 1671 A., ... Stuart, S. N. (2023). Ongoing declines for the world’s amphibians in the face of
 1672 emerging threats. *Nature*, 622(7982), 308–314. [https://doi.org/10.1038/s41586-023-](https://doi.org/10.1038/s41586-023-06578-4)
 1673 06578-4

1674 Lundmark, H., Josefsson, T., & Östlund, L. (2017). The introduction of modern forest
 1675 management and clear-cutting in Sweden: Ridö State Forest 1832–2014. *European*
 1676 *Journal of Forest Research*, 136(2), 269–285. <https://doi.org/10.1007/s10342-017-1027-6>

1677 Máliš, F., Ujházy, K., Hederová, L., Ujházyová, M., Csölleová, L., Coomes, D. A., & Zellweger,
 1678 F. (2023). Microclimate variation and recovery time in managed and old-growth

temperate forests. *Agricultural and Forest Meteorology*, 342, 109722.
<https://doi.org/10.1016/j.agrformet.2023.109722>

Martin, T. L., & Huey, R. B. (2008). Why “Suboptimal” Is Optimal: Jensen’s Inequality and Ectotherm Thermal Preferences. *The American Naturalist*, 171(3), E102–E118.
<https://doi.org/10.1086/527502>

Matlack, G. R. (1993). Microenvironment variation within and among forest edge sites in the eastern United States. *Biological Conservation*, 66(3), 185–194.
[https://doi.org/10.1016/0006-3207\(93\)90004-K](https://doi.org/10.1016/0006-3207(93)90004-K)

Mazerolle, M. J., Lapointe, M. S. P., Imbeau, L., & Joanisse, G. (2021). Woodland salamander population structure and body condition under irregular shelterwood systems. *Canadian Journal of Forest Research*, 51(9), 1281–1291. <https://doi.org/10.1139/cjfr-2020-0405>

Mitchell, A., & Bergmann, P. J. (2016). Thermal and moisture habitat preferences do not maximize jumping performance in frogs. *Functional Ecology*, 30(5), 733–742.
<https://doi.org/10.1111/1365-2435.12535>

Mitchell, W. A., & Angilletta Jr, M. J. (2009). Thermal games: Frequency-dependent models of thermal adaptation. *Functional Ecology*, 23(3), 510–520. <https://doi.org/10.1111/j.1365-2435.2009.01542.x>

Moore, F. R., & Gatten Jr, R. E. (1989). Locomotor performance of hydrated, dehydrated, and osmotically stressed anuran amphibians. *Herpetologica*, 1, 101–110.

Murcia, C. (1995). Edge effects in fragmented forests: Implications for conservation. *Trends in Ecology & Evolution*, 10(2), 58–62. [https://doi.org/10.1016/S0169-5347\(00\)88977-6](https://doi.org/10.1016/S0169-5347(00)88977-6)

1700 Natural Resources, O. M. & Forestry. (2001). *Forest management guide for natural disturbance*
 1701 *pattern emulation, Version 3.1*. Ontario Ministry of Natural Resources. Queen's Printer
 1702 for Ontario, Toronto.

1703 Natural Resources, O. M. & Forestry. (2023). *Forest Management Guide for Boreal Landscapes*
 1704 *in Ontario*.

1705 Navas, C. A. (1996). Metabolic Physiology, Locomotor Performance, and Thermal Niche
 1706 Breadth in Neotropical Anurans. *Physiological Zoology*, 69(6), 1481–1501.
 1707 <https://doi.org/10.1086/physzool.69.6.30164271>

1708 Navas, C. A., Gomes, F. R., & Carvalho, J. E. (2008). Thermal relationships and exercise
 1709 physiology in anuran amphibians: Integration and evolutionary implications.
 1710 *Comparative Biochemistry and Physiology Part A: Molecular & Integrative Physiology*,
 1711 151(3), 344–362. <https://doi.org/10.1016/j.cbpa.2007.07.003>

1712 Navas, C. A., Gomes, F. R., & De Domenico, E. A. (2016). Physiological ecology and
 1713 conservation of anuran amphibians. In D. V. Andrade, C. R. Bevier, & J. E. Carvalho
 1714 (Eds.), *Amphibian and Reptile Adaptations to the Environment: Interplay Between*
 1715 *Physiology and Behavior*. (pp. 155–188). CRC Press.

1716 Navas, C. A., Gouveia, S. F., Solano-Iguarán, J. J., Vidal, M. A., & Bacigalupe, L. D. (2021).
 1717 Amphibian responses in experimental thermal gradients: Concepts and limits for
 1718 inference. *Comparative Biochemistry and Physiology Part B: Biochemistry and*
 1719 *Molecular Biology*, 254, 110576. <https://doi.org/10.1016/j.cbpb.2021.110576>

1720 Neel, L. K., & McBrayer, L. D. (2018). Habitat management alters thermal opportunity.
 1721 *Functional Ecology*, 32(8), 2029–2039. <https://doi.org/10.1111/1365-2435.13123>

1722 Nowakowski, A. J., Watling, J. I., Thompson, M. E., Brusch, G. A., Catenazzi, A., Whitfield, S.
 1723 M., Kurz, D. J., Suárez-Mayorga, Á., Aponte-Gutiérrez, A., Donnelly, M. A., & Todd, B.
 1724 D. (2018). Thermal biology mediates responses of amphibians and reptiles to habitat
 1725 modification. *Ecology Letters*, 21(3), 345–355. <https://doi.org/10.1111/ele.12901>
 1726 Olson, E. R., Laughlin, M. M., & Martin, J. G. (2023). Reaching New Heights: Arboreal Camera
 1727 Trapping Provides New Insights on the Ecology of Gray Treefrogs (*Hyla versicolor*).
 1728 *Journal of Herpetology*, 57(3), 281–289. <https://doi.org/10.1670/22-060>
 1729 Padfield, D., O’Sullivan, H., & Pawar, S. (2021). rTPC and nls.multstart: A new pipeline to fit
 1730 thermal performance curves in r. *Methods in Ecology and Evolution*, 12(6), 1138–1143.
 1731 <https://doi.org/10.1111/2041-210X.13585>
 1732 Patrick, D. A., Hunter, M. L., & Calhoun, A. J. K. (2006). Effects of experimental forestry
 1733 treatments on a Maine amphibian community. *Forest Ecology and Management*, 234(1),
 1734 323–332. <https://doi.org/10.1016/j.foreco.2006.07.015>
 1735 Payne, N. L., Smith, J. A., Meulen, D. E., Taylor, M. D., Watanabe, Y. Y., Takahashi, A.,
 1736 Marzullo, T. A., Gray, C. A., Cadiou, G., & Suthers, I. M. (2015). Temperature
 1737 dependence of fish performance in the wild: Links with species biogeography and
 1738 physiological thermal tolerance. *Functional Ecology*, 30(6), 903–912.
 1739 <https://doi.org/10.1111/1365-2435.12618>
 1740 Perez-Navarro, M. A., Lloret, F., Molina-Venegas, R., Alcántara, J. M., & Verdú, M. (2024).
 1741 Plant canopies promote climatic disequilibrium in Mediterranean recruit communities.
 1742 *Ecology Letters*, 27(2), e14391. <https://doi.org/10.1111/ele.14391>

1743 Peterman, W. E., Locke, J. L., & Semlitsch, R. D. (2013). Spatial and temporal patterns of water
 1744 loss in heterogeneous landscapes: Using plaster models as amphibian analogues.
 1745 *Canadian Journal of Zoology*, 91(3), 135–140. <https://doi.org/10.1139/cjz-2012-0229>
 1746 Peterman, W. E., & Semlitsch, R. D. (2014). Spatial variation in water loss predicts terrestrial
 1747 salamander distribution and population dynamics. *Oecologia*, 176(2), 357–369.
 1748 <https://doi.org/10.1007/s00442-014-3041-4>
 1749 Pinheiro, J. P., Bates, D., DebRoy, S., Sarkar, D., Heisterkamp, S., Van Willigen, B., Ranke, J.,
 1750 & R Core Team. (2025). *nlme: Linear and Nonlinear Mixed Effects Models* (Version 3.1-
 1751 168) [Computer software]. <https://cran.r-project.org/web/packages/nlme/index.html>
 1752 Popescu, V. D., & Hunter, M. L. (2011). Clear-cutting affects habitat connectivity for a forest
 1753 amphibian by decreasing permeability to juvenile movements. *Ecological Applications*,
 1754 21(4), 1283–1295. <https://doi.org/10.1890/10-0658.1>
 1755 Pottier, P., Burke, S., Drobniak, S. M., Lagisz, M., & Nakagawa, S. (2021). Sexual (in)equality?
 1756 A meta-analysis of sex differences in thermal acclimation capacity across ectotherms.
 1757 *Functional Ecology*, 35(12), 2663–2678. <https://doi.org/10.1111/1365-2435.13899>
 1758 Pottier, P., Kearney, M. R., Wu, N. C., Gunderson, A. R., Rej, J. E., Rivera-Villanueva, A. N.,
 1759 Pollo, P., Burke, S., Drobniak, S. M., & Nakagawa, S. (2025). Vulnerability of
 1760 amphibians to global warming. *Nature*, 639(8056), 954–961.
 1761 <https://doi.org/10.1038/s41586-025-08665-0>
 1762 Pough, F. H., Taigen, T. L., Stewart, M. M., & Brussard, P. F. (1983). Behavioral Modification
 1763 of Evaporative Water Loss by a Puerto Rican Frog. *Ecology*, 64(2), 244–252.
 1764 <https://doi.org/10.2307/1937072>

1765 Prates, I., & Navas, C. A. (2009). Cutaneous Resistance to Evaporative Water Loss in Brazilian
 1766 Rhinella (Anura: Bufonidae) from Contrasting Environments. *Copeia*, 2009(3), 618–622.
 1767 <https://doi.org/10.1643/CP-08-128>

1768 Preest, M. R., & Pough, F. H. (1989). Interaction of Temperature and Hydration on Locomotion
 1769 of Toads. *Functional Ecology*, 3(6), 693. <https://doi.org/10.2307/2389501>

1770 Price, D. T. (2013). Anticipating the consequences of climate change for Canada’s boreal forest
 1771 ecosystems. *Environmental Reviews*, 21, 322–365.

1772 Ratkowsky, D. A., Lowry, R. K., McMeekin, T. A., Stokes, A. N., & Chandler, R. E. (1983).
 1773 Model for bacterial culture growth rate throughout the entire biokinetic temperature
 1774 range. *Journal of Bacteriology*, 154(3), 1222–1226.
 1775 <https://doi.org/10.1128/jb.154.3.1222-1226.1983>

1776 Remmel, T. K., Ouellette, M., & Wu, W. J. (2023). A boreal wildfire and harvesting database
 1777 with ensemble confidence attributes for Ontario (1972–2021+). *International Journal of*
 1778 *Applied Earth Observation and Geoinformation*, 117, 103199.
 1779 <https://doi.org/10.1016/j.jag.2023.103199>

1780 Rezende, E. L., & Bozinovic, F. (2019). Thermal performance across levels of biological
 1781 organization. *Philosophical Transactions of the Royal Society of London. Series B,*
 1782 *Biological Sciences*, 374(1778), 20180549. <https://doi.org/10.1098/rstb.2018.0549>

1783 Richard, B., Dupouey, J.-L., Corcket, E., Alard, D., Archaux, F., Aubert, M., Boulanger, V.,
 1784 Gillet, F., Langlois, E., Macé, S., Montpied, P., Beaufiles, T., Begeot, C., Behr, P.,
 1785 Boissier, J.-M., Camaret, S., Chevalier, R., Decocq, G., Dumas, Y., ... Lenoir, J. (2021).
 1786 The climatic debt is growing in the understorey of temperate forests: Stand characteristics

1787 matter. *Global Ecology and Biogeography*, 30(7), 1474–1487.

1788 <https://doi.org/10.1111/geb.13312>

1789 Rittenhouse, T. A. G., Harper, E. B., Rehard, L. R., & Semlitsch, R. D. (2008). The Role of

1790 Microhabitats in the Desiccation and Survival of Anurans in Recently Harvested Oak–

1791 Hickory Forest. *Copeia*, 2008(4), 807–814. <https://doi.org/10.1643/CH-07-176>

1792 Rittenhouse, T. A. G., Semlitsch, R. D., & Thompson, F. R. I. (2009). Survival costs associated

1793 with wood frog breeding migrations: Effects of timber harvest and drought. *Ecology*,

1794 90(6): 1620–1630., 90, 1620–1630. <https://doi.org/10.1890/08-0326.1>

1795 Robinson, M., Nielsen, S. E., Eaton, B., & Paszkowski, C. (2023). Effect of variable retention

1796 forestry on wood frogs (*Lithobates sylvaticus*) in early successional boreal mixedwood

1797 forests. *Canadian Journal of Zoology*, 101(3), 172–188. [https://doi.org/10.1139/cjz-2022-](https://doi.org/10.1139/cjz-2022-0062)

1798 0062

1799 Rogowitz, G. L., Cortés-Rivera, M., & Nieves-Puigdollér, K. (1999). Water loss, cutaneous

1800 resistance, and effects of dehydration on locomotion of Eleutherodactylus frogs. *Journal*

1801 *of Comparative Physiology B: Biochemical, Systemic, and Environmental Physiology*,

1802 169(3), 179–186. <https://doi.org/10.1007/s003600050209>

1803 Rohr, J. R., Civitello, D. J., Cohen, J. M., Roznik, E. A., Sinervo, B., & Dell, A. I. (2018). The

1804 complex drivers of thermal acclimation and breadth in ectotherms. *Ecology Letters*,

1805 21(9), 1425–1439. <https://doi.org/10.1111/ele.13107>

1806 Rollins-Smith, L. A., & Le Sage, E. H. (2023). Heat stress and amphibian immunity in a time of

1807 climate change. *Philosophical Transactions of the Royal Society B: Biological Sciences*,

1808 378(1882), 20220132. <https://doi.org/10.1098/rstb.2022.0132>

1809 Rome, L. C., Stevens, E. D., & John-Alder, H. B. (1992). Influence of Temperature and Thermal
1810 Acclimation on Physiological Function. In M. E. Bicudo & W. W. Burggren (Eds.),
1811 *Environmental Physiology of the Amphibians*. The University of Chicago Press.

1812 Rozen-Rechels, D., Dupoué, A., Lourdais, O., Chamaillé-Jammes, S., Meylan, S., Clobert, J., &
1813 Le Galliard, J.-F. (2019). When water interacts with temperature: Ecological and
1814 evolutionary implications of thermo-hydroregulation in terrestrial ectotherms. *Ecology*
1815 *and Evolution*, 9(17), 10029–10043. <https://doi.org/10.1002/ece3.5440>

1816 Roznik, E. A., Rodriguez-Barbosa, C. A., & Johnson, S. A. (2018). Hydric Balance and
1817 Locomotor Performance of Native and Invasive Frogs. *Frontiers in Ecology and*
1818 *Evolution*, 6, 159. <https://doi.org/10.3389/fevo.2018.00159>

1819 Ruiz, T., Koussoroplis, A.-M., Danger, M., Aguer, J.-P., Morel-Desrosiers, N., & Bec, A.
1820 (2019). *The Threshold Elemental Ratio of an ectotherm decreases then increases with*
1821 *rising temperature* (p. 681239). bioRxiv. <https://doi.org/10.1101/681239>

1822 Rutschmann, A., Perry, C., Le Galliard, J.-F., Dupoué, A., Lourdais, O., Guillon, M., Brusch IV,
1823 G., Cote, J., Richard, M., Clobert, J., & Miles, D. B. (2024). Ecological responses of
1824 squamate reptiles to nocturnal warming. *Biological Reviews*, 99(2), 598–621.
1825 <https://doi.org/10.1111/brv.13037>

1826 Scheffers, B. R., Brunner, R. M., Ramirez, S. D., Shoo, L. P., Diesmos, A., & Williams, S. E.
1827 (2013). Thermal Buffering of Microhabitats is a Critical Factor Mediating Warming
1828 Vulnerability of Frogs in the Philippine Biodiversity Hotspot. *Biotropica*, 45(5), 628–
1829 635. <https://doi.org/10.1111/btp.12042>

1830 Schmidt, R., Zummach, C., Sinai, N., Sabino-Pinto, J., Künzel, S., Dausmann, K. H., &
1831 Ruthsatz, K. (2024). Physiological responses to a changing winter climate in an early

1832 spring-breeding amphibian. *Ecology and Evolution*, 14(7), e70042.

1833 <https://doi.org/10.1002/ece3.70042>

1834 Schulte, P. M., Healy, T. M., & Fangue, N. A. (2011). Thermal Performance Curves, Phenotypic

1835 Plasticity, and the Time Scales of Temperature Exposure. *Integrative and Comparative*

1836 *Biology*, 51(5), 691–702. <https://doi.org/10.1093/icb/icr097>

1837 Sears, M. W., & Angilletta, M. J. (2015). Costs and benefits of thermoregulation revisited: Both

1838 the heterogeneity and spatial structure of temperature drive energetic costs. *The American*

1839 *Naturalist*, 185:E94–E102.

1840 Sears, M. W., Angilletta, M. J., Schuler, M. S., Borchert, J., Dilliplane, K. F., & Stegman, M.

1841 (2016). Configuration of the thermal landscape determines thermoregulatory performance

1842 of ectotherms. *Proceedings of the National Academy of Sciences of the United States of*

1843 *America*, 113, 10595–10600.

1844 Semlitsch, R. (2000). Principles for Management of Aquatic-Breeding Amphibians. *The Journal*

1845 *of Wildlife Management*, 64, 615. <https://doi.org/10.2307/3802732>

1846 Semlitsch, R. D., Todd, B. D., Blomquist, S. M., Calhoun, A. J. K., Gibbons, J. W., Gibbs, J. P.,

1847 Graeter, G. J., Harper, E. B., Hocking, D. J., Hunter, M. L., Jr., Patrick, D. A.,

1848 Rittenhouse, T. A. G., & Rothermel, B. B. (2009). Effects of Timber Harvest on

1849 Amphibian Populations: Understanding Mechanisms from Forest Experiments.

1850 *BioScience*, 59(10), 853–862. <https://doi.org/10.1525/bio.2009.59.10.7>

1851 Senzano, L. M., & Andrade, D. V. (2018). Temperature and dehydration effects on metabolism,

1852 water uptake, and the partitioning between respiratory and cutaneous evaporative water

1853 loss in a terrestrial toad. *Journal of Experimental Biology*, jeb.188482.

1854 <https://doi.org/10.1242/jeb.188482>

1855 Shi, P., & Ge, F. (2010). A comparison of different thermal performance functions describing
1856 temperature-dependent development rates. *Journal of Thermal Biology*, 35(5), 225–231.
1857 <https://doi.org/10.1016/j.jtherbio.2010.05.005>

1858 Shoemaker, V. H., Hillman, S. S., Hillyard, S. D., Jackson, D. C., Mcclanahan, L. L., Withers,
1859 P., & Wygoda, M. (1992). Exchange of water, ions and respiratory gases in terrestrial
1860 amphibians. In M. Feder & W. Burggren (Eds.), *Environmental Physiology of the*
1861 *Amphibians* (pp. 125–150). The University of Chicago Press.

1862 Shoemaker, V. H., & Nagy, K. A. (1977). Osmoregulation in amphibians and reptiles. *Annual*
1863 *Review of Physiology*, 39, 449–471.
1864 <https://doi.org/10.1146/annurev.ph.39.030177.002313>

1865 Sinervo, B., Méndez-de-la-Cruz, F., Miles, D. B., Heulin, B., Bastiaans, E., Villagrán-Santa
1866 Cruz, M., Lara-Resendiz, R., Martínez-Méndez, N., Calderón-Espinosa, M. L., Meza-
1867 Lázaro, R. N., Gadsden, H., Avila, L. J., Morando, M., De la Riva, I. J., Sepulveda, P. V.,
1868 Rocha, C. F. D., Ibarguengoytia, N., Puntriano, C. A., Massot, M., ... Sites, J. W. (2010).
1869 Erosion of Lizard Diversity by Climate Change and Altered Thermal Niches. *Science*,
1870 328(5980), 894–899. <https://doi.org/10.1126/science.1184695>

1871 Singmann, H., Bolker, B., Westfall, J., Aust, F., Ben-Shachar, M. S., Højsgaard, S., Fox, J.,
1872 Lawrence, M. A., Mertens, U., Love, J., Lenth, R., & Christensen, R. H. B. (2024). *afex*:
1873 *Analysis of Factorial Experiments* (Version 1.4-1) [Computer software]. [https://cran.r-](https://cran.r-project.org/web/packages/afex/index.html)
1874 [project.org/web/packages/afex/index.html](https://cran.r-project.org/web/packages/afex/index.html)

1875 Soltani, K., Amiri, A., Ebtehaj, I., Cheshmehghasabani, H., Fazeli, S., Gumiere, S. J., &
1876 Bonakdari, H. (2024). Advanced Forecasting of Drought Zones in Canada Using Deep

1877 Learning and CMIP6 Projections. *Climate*, 12(8), Article 8.

1878 <https://doi.org/10.3390/cli12080119>

1879 Spotila, J. R., O'Connor, M. P., & Bakken, G. S. (1992). Biophysics of heat and mass transfer.

1880 Pages 59-81 in M.E. In Bicudo & W. W. Burggren (Eds.), *Environmental Physiology of*

1881 *the Amphibians*. The University of Chicago Press.

1882 Stuart, S. N., Chanson, J. S., Cox, N. A., Young, B. E., Rodrigues, A. S. L., Fischman, D. L., &

1883 Waller, R. W. (2004). Status and Trends of Amphibian Declines and Extinctions

1884 Worldwide. *Science*, 306(5702), 1783–1786. <https://doi.org/10.1126/science.1103538>

1885 Sunday, J. M., Bates, A. E., Kearney, M. R., Colwell, R. K., Dulvy, N. K., Longino, J. T., &

1886 Huey, R. B. (2014). Thermal-safety margins and the necessity of thermoregulatory

1887 behavior across latitude and elevation. *Proceedings of the National Academy of Sciences*,

1888 111(15), 5610–5615. <https://doi.org/10.1073/pnas.1316145111>

1889 Taylor, S. E., & Sexton, O. J. (1972). Some Implications of Leaf Tearing in Musaceae. *Ecology*,

1890 53(1), 143–149. <https://doi.org/10.2307/1935720>

1891 Thomas, M. K., Kremer, C. T., Klausmeier, C. A., & Litchman, E. (2012). A global pattern of

1892 thermal adaptation in marine phytoplankton. *Science (New York, N.Y.)*, 338(6110), 1085–

1893 1088. <https://doi.org/10.1126/science.1224836>

1894 Todd, B. D., Blomquist, S. M., Harper, E. B., & Osbourn, M. S. (2014). Effects of timber

1895 harvesting on terrestrial survival of pond-breeding amphibians. *Forest Ecology and*

1896 *Management*, 313, 123–131. <https://doi.org/10.1016/j.foreco.2013.11.011>

1897 Todd, B. D., Luhring, T. M., Rothermel, B. B., & Gibbons, J. W. (2009). Effects of forest

1898 removal on amphibian migrations: Implications for habitat and landscape connectivity.

1899 *Journal of Applied Ecology*, 46(3), 554–561. <https://doi.org/10.1111/j.1365->

1900 2664.2009.01645.x

1901 Todd, B. D., & Rothermel, B. B. (2006). Assessing quality of clearcut habitats for amphibians:

1902 Effects on abundances versus vital rates in the southern toad (*Bufo terrestris*). *Biological*

1903 *Conservation*, 133(2), 178–185. <https://doi.org/10.1016/j.biocon.2006.06.003>

1904 Toledo, R. C., & Jared, C. (1993). Cutaneous adaptations to water balance in amphibians.

1905 *Comparative Biochemistry and Physiology Part A: Physiology*, 105(4), 593–608.

1906 [https://doi.org/10.1016/0300-9629\(93\)90259-7](https://doi.org/10.1016/0300-9629(93)90259-7)

1907 Tomlinson, S., & Phillips, R. D. (2015). Differences in metabolic rate and evaporative water loss

1908 associated with sexual dimorphism in thynnine wasps. *Journal of Insect Physiology*, 78,

1909 62–68. <https://doi.org/10.1016/j.jinsphys.2015.04.011>

1910 Tracy, C. R. (1976). A Model of the Dynamic Exchanges of Water and Energy between a

1911 Terrestrial Amphibian and Its Environment. *Ecological Monographs*, 46(3), 293–326.

1912 <https://doi.org/10.2307/1942256>

1913 Tracy, C. R., Betts, G., Richard Tracy, C., & Christian, K. A. (2007). Plaster Models to Measure

1914 Operative Temperature and Evaporative Water Loss of Amphibians. *Journal of*

1915 *Herpetology*, 41(4), 597–603. <https://doi.org/10.1670/07-006.1>

1916 Tracy, C. R., Christian, K. A., & Tracy, C. R. (2010). Not just small, wet, and cold: Effects of

1917 body size and skin resistance on thermoregulation and arboreality of frogs. *Ecology*,

1918 91(5), 1477–1484. <https://doi.org/10.1890/09-0839.1>

1919 Tracy, C. R., Tixier, T., Le Nöene, C., & Christian, K. A. (2014). Field Hydration State Varies

1920 among Tropical Frog Species with Different Habitat Use. *Physiological and Biochemical*

1921 *Zoology*, 87(2), 197–202. <https://doi.org/10.1086/674537>

1922 Tuff, K. T., Tuff, T., & Davies, K. F. (2016). A framework for integrating thermal biology into
 1923 fragmentation research. *Ecology Letters*, 19(4), 361–374.
 1924 <https://doi.org/10.1111/ele.12579>

1925 Urban, M. C., Tewksbury, J. J., & Sheldon, K. S. (2012). On a collision course: Competition and
 1926 dispersal differences create no-analogue communities and cause extinctions during
 1927 climate change. *Proceedings. Biological Sciences*, 279(1735), 2072–2080.
 1928 <https://doi.org/10.1098/rspb.2011.2367>

1929 Veysey Powell, J. S., & Babbitt, K. J. (2015). Despite Buffers, Experimental Forest Clearcuts
 1930 Impact Amphibian Body Size and Biomass. *PLoS ONE*, 10(11), e0143505.
 1931 <https://doi.org/10.1371/journal.pone.0143505>

1932 Weber, M. G., & Flannigan, M. D. (1997). Canadian boreal forest ecosystem structure and
 1933 function in a changing climate: Impact on fire regimes. *Environmental Reviews*, 5(3–4),
 1934 145–166. <https://doi.org/10.1139/a97-008>

1935 Wells, K. D. (2010). *The Ecology and Behavior of Amphibians*. University of Chicago Press.
 1936 <https://press.uchicago.edu/ucp/books/book/chicago/E/bo5298950.html>

1937 Wilson, R. S. (2001). Geographic variation in thermal sensitivity of jumping performance in the
 1938 frog *Limnodynastes peronii*. *Journal of Experimental Biology*, 204(24), 4227–4236.
 1939 <https://doi.org/10.1242/jeb.204.24.4227>

1940 Wolf, K. D., Higuera, P. E., Davis, K. T., & Dobrowski, S. Z. (2021). Wildfire impacts on forest
 1941 microclimate vary with biophysical context. *Ecosphere*, 12(5), e03467.
 1942 <https://doi.org/10.1002/ecs2.3467>

1943 Wulder, M. A., Hermosilla, T., White, J. C., Bater, C. W., Hobart, G., & Bronson, S. C. (2024).
 1944 Development and implementation of a stand-level satellite-based forest inventory for

1945 Canada. *Forestry: An International Journal of Forest Research*, 97(4), 546–563.

1946 <https://doi.org/10.1093/forestry/cpad065>

1947 Young, J. E., Christian, K. A., Donnellan, S., Tracy, C. R., & Parry, D. (2005). Comparative

1948 Analysis of Cutaneous Evaporative Water Loss in Frogs Demonstrates Correlation with

1949 Ecological Habits. *Physiological and Biochemical Zoology*, 78(5), 847–856.

1950 <https://doi.org/10.1086/432152>

1951 Zhang, F., Biederman, J. A., Dannenberg, M. P., Yan, D., Reed, S. C., & Smith, W. K. (2021).

1952 Five Decades of Observed Daily Precipitation Reveal Longer and More Variable Drought

1953 Events Across Much of the Western United States. *Geophysical Research Letters*, 48(7),

1954 e2020GL092293. <https://doi.org/10.1029/2020GL092293>

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1970 **7 APPENDIX A**

1971 **7.1 Weighted average results for individual thermal performance curves**

1972
1973 **Table 7.1** Delta AIC values for each model and frog included in the final thermal performance
1974 curve at 100% hydration. Model name indicates all models fit for each frog. NAs in column
1975 values indicates that the specified model did not fit for the corresponding frog. Models with
1976 lower delta AIC were given higher weights towards the final curves for each frog.
1977

Model Name	Δ AIC								
	A1	A2	A3	A5	A6	B3	B4	B5	B6
Analytis K I	2.13	5.76	4.77	7.83	9.05	5.08	1.69	6.07	0.30
Ashrafi I	1.10	0.00	0.70	0.01	0.00	3.68	2.55	NA	1.03
Atkin	0.90	0.78	1.53	2.12	5.05	3.87	0.37	0.00	0.23
Briere I	2.94	2.17	2.16	0.91	7.21	5.35	3.93	3.94	3.33
Briere I no int	7.30	11.20	12.34	14.03	12.16	5.51	12.77	4.01	5.95
Eubank	0.93	0.56	1.46	0.00	5.77	3.38	1.67	0.21	0.38
Gaussian	0.91	1.05	1.64	2.17	6.51	3.98	0.00	2.73	0.00
Gaussian Gomp	2.80	2.85	7.63	4.10	11.23	5.76	2.27	NA	2.00
Janisch I	0.92	0.65	1.50	1.40	5.98	3.68	1.01	0.81	0.19
Lactin I	1.34	2.08	2.45	4.73	NA	3.51	2.38	2.47	1.51
Lactin II	2.87	2.74	2.92	NA	NA	6.82	4.38	10.45	3.57
Logan I	2.86	2.74	NA	NA	9.28	5.61	4.24	NA	3.50
Ratkowsky	4.98	7.02	8.07	NA	NA	NA	5.63	NA	NA
Rezende	2.87	2.26	2.79	NA	11.71	5.51	3.59	6.62	3.11
Ruiz	2.90	2.05	0.00	NA	NA	NA	2.00	3.15	2.00
Simplified Beta	NA	NA	NA	NA	NA	NA	NA	NA	0.08
Taylor S	2.45	3.27	5.98	8.79	8.63	4.06	1.42	2.64	0.48
Tomlinson P	0.00	1.65	0.82	6.91	NA	0.00	5.62	4.39	3.66
Weibull	2.06	2.87	2.84	2.05	8.36	NA	0.33	3.98	1.75

1978

1979

1980 **Table 7.2.** Delta AIC values for individually fit thermal performance curves at 92% hydration.
 1981 Model name indicates all models fit for each frog. NAs in column values indicates that the
 1982 specified model did not fit for the corresponding frog. Models with lower delta AIC were given
 1983 higher weights towards the final curves for each frog.

Model Name	Δ AIC								
	A1	A2	A3	A5	A6	B3	B4	B5	B6
Analytis K I	5.67	0.65	3.78	1.34	3.43	4.13	2.34	8.49	6.55
Ashrafi	1.03	5.18	2.78	NA	0.00	4.57	1.34	1.53	5.64
Atkin	NA	2.57	2.49	0.00	NA	4.39	2.22	NA	4.23
Briere I	3.35	6.30	3.95	2.40	26.27	6.43	2.44	3.49	6.14
Briere II	12.19	0.00	4.78	2.94	10.33	5.36	5.75	17.12	6.08
Eubank	0.18	4.96	0.00	NA	22.91	4.64	1.82	0.34	0.26
Gaussian	1.92	0.32	3.05	0.76	28.75	4.16	2.43	2.31	5.40
Gaussian Gomp	14.63	2.32	NA	3.21	NA	6.10	4.55	3.71	10.82
Janisch I	0.84	3.52	1.48	0.15	22.13	4.47	2.09	1.13	2.78
Lactin I	3.15	6.09	1.16	0.42	NA	4.79	1.30	5.60	2.07
Lactin II	2.77	8.17	NA	NA	10.23	6.86	1.35	3.20	NA
Logan I	NA	NA	NA	NA	NA	NA	4.76	15.38	NA
Mitchell angilletta	7.87	NA	3.87	NA	18.85	NA	NA	6.93	6.88
Rezende	NA	NA	3.33	NA	28.03	8.17	3.46	NA	4.30
Simplified Beta	2.46	NA	4.25	NA	10.39	3.81	3.04	NA	6.47
Taylor S	5.34	2.12	3.44	NA	NA	4.30	2.18	10.35	5.90
Tomlinson P	0.00	8.51	NA	0.00	14.95	5.11	0.00	0.00	0.00
Weibull	4.04	NA	4.47	2.53	17.87	0.00	3.99	4.04	5.67

1984

1985

1986

1987 **Table 7.3.** Delta AIC values for individually fit thermal performance curves at 84% hydration.
 1988 Model name indicates all models fit for each frog. NAs in column values indicates that the
 1989 specified model did not fit for the corresponding frog. Models with lower delta AIC were given
 1990 higher weights towards the final curves for each frog.

Model Name	Δ AIC								
	A1	A2	A3	A5	A6	B3	B4	B5	B6
Analytis K I	3.88	3.45	1.07	0.87	9.11	7.66	7.87	5.55	5.74
Ashrafi	1.98	4.35	0.32	NA	NA	0.84	5.91	0.34	6.45
Atkin	3.79	2.73	0.74	0.10	NA	0.00	7.22	0.00	5.41
Briere I	1.79	7.92	1.91	NA	1.08	9.32	3.63	2.33	NA
Eubank	3.44	0.00	0.60	0.00	NA	NA	6.22	0.31	4.47
Gaussian	4.06	4.19	1.13	0.95	9.43	6.47	11.38	2.43	5.80
Gaussian Gomp	6.37	3.80	2.68	2.44	12.47	NA	NA	2.02	6.94
Janisch I	3.69	0.90	0.68	0.36	2.27	NA	7.28	4.68	4.71
Janisch II	6.89	NA	2.55	2.10	6.55	NA	NA	1.72	NA
Lactin II	2.50	6.20	NA	NA	4.86	NA	NA	2.65	8.31
Logan I	6.84	NA	2.63	2.54	11.99	NA	NA	2.09	8.26
Mitchell A	NA	2.14	NA	NA	7.06	5.75	NA	3.51	5.02
Rezende Bozinovic	4.77	NA	2.67	NA	20.88	2.82	6.34	NA	NA
Ruiz	NA	NA	NA	0.67	NA	2.48	NA	NA	NA
Simplified Beta	5.06	2.21	1.14	NA	4.89	NA	9.06	0.00	4.80
Simplified Briere	3.94	3.86	0.95	0.49	5.39	10.28	8.26	3.09	5.92
Taylor S	4.22	3.96	1.16	0.69	10.57	NA	7.21	6.87	6.15
Tomlinson P	0.00	5.26	0.00	0.10	0.00	NA	0.00	0.78	NA
Weibull	5.44	NA	2.73	2.26	5.09	4.96	8.56	2.00	0.00

1991

1992

1993 **Table 7.4.** Delta AIC values for individually fit thermal performance curves at 75% hydration.
1994 Model name indicates all models fit for each frog. NAs in column values indicates that the
1995 specified model did not fit for the corresponding frog. Models with lower delta AIC were given
1996 higher weights towards the final curves for each frog.

Model Name	Δ AIC								
	A1	A2	A3	A5	A6	B3	B4	B5	B6
Analytis K I	1.59	5.37	4.07	0.40	NA	0.90	11.17	1.95	0.78
Ashrafi I	0.00	NA	3.10	0.00	NA	0.57	4.66	NA	1.45
Atkin I	0.48	0.24	3.20	0.10	NA	0.86	NA	0.00	1.42
Briere I	2.43	NA	3.23	1.99	NA	2.13	NA	NA	3.62
Eubank	0.37	0.00	1.67	0.11	NA	0.65	0.00	0.59	1.02
Gaussian	0.84	1.52	5.48	0.43	4.78	0.92	10.31	1.52	4.19
Gaussian Gomp	NA	2.79	NA	2.16	NA	3.05	NA	2.21	2.05
Janisch I	0.51	1.56	2.96	1.01	NA	2.79	7.35	0.99	2.30
Lactin II	NA	NA	NA	NA	NA	NA	NA	NA	4.80
Logan I	NA	NA	9.22	2.10	NA	2.80	NA	2.08	NA
Rezende	2.98	NA	3.68	2.83	NA	2.58	20.77	6.57	6.25
Simplified Beta	2.12	NA	4.54	0.70	NA	1.29	5.33	NA	0.45
Simplified Briere	1.85	4.90	4.57	0.25	NA	0.88	14.96	1.06	0.00
Simplified Briere II	3.32	2.97	NA	NA	NA	NA	NA	2.90	NA
Taylor S	NA	NA	3.65	NA	0.62	NA	NA	NA	2.59
Tomlinson P	0.47	NA	0.00	0.05	NA	0.00	13.21	NA	3.40
Weibull	2.66	2.24	4.80	2.16	0.00	2.81	NA	2.00	NA

1997
1998
1999
2000

2001 **Table 7.5.** Delta AIC values for individually fit thermal performance curves at 70% hydration.
 2002 Model name indicates all models fit for each frog. NAs in column values indicates that the
 2003 specified model did not fit for the corresponding frog. Models with lower delta AIC were given
 2004 higher weights towards the final curves for each frog.

Model Name	Δ AIC							
	A1	A2	A3	A5	B3	B4	B5	B6
Analytis K I	0.85	0.56	2.70	0.85	0.17	3.65	1.50	2.42
Ashrafi	0.86	NA	3.55	NA	0.02	NA	1.89	2.72
Ashrafi II	0.81	NA	4.40	NA	0.04	NA	0.34	0.00
Atkin	0.83	0.00	2.40	0.01	0.00	1.00	1.77	3.57
Briere I	1.04	0.84	2.91	1.23	0.19	7.04	1.94	3.08
Eubank	0.00	0.07	0.00	0.00	0.04	4.33	NA	4.52
Gaussian	0.88	0.47	4.55	0.88	0.16	3.65	2.81	5.00
Gaussian Gomp	NA	NA	6.58	NA	NA	3.71	2.97	5.01
Janisch I	0.23	1.23	0.61	0.90	0.24	9.80	0.43	NA
Lactin I	0.85	NA	NA	0.28	0.05	NA	NA	NA
Lactin II	NA	NA	NA	NA	NA	NA	NA	6.44
Logan I	NA	NA	6.66	NA	NA	4.55	NA	NA
Mitchell A	NA	NA	NA	NA	NA	0.00	NA	NA
Simplified Beta	1.44	1.93	4.26	1.14	0.56	5.74	1.26	9.72
Simplified Briere	NA	0.55	4.22	0.70	0.06	2.67	1.08	0.57
Taylor S	0.85	0.44	NA	0.77	0.16	4.23	1.76	2.88
Tomlinson phil	0.83	0.00	NA	0.01	0.00	1.01	2.88	4.80
Weibull	2.20	2.29	5.64	2.36	1.68	3.40	0.00	3.04

2005

2006 7.2 Weighted average results for individual hydro-performance curves.

2007 **Table 7.6** Delta AIC values for individually fit hydro-performance curves at 33 °C. Model name
 2008 indicates all models fit for each frog. NAs in column values indicates that the specified model
 2009 did not fit for the corresponding frog. Models with lower delta AIC were given higher weights
 2010 towards the final curves for each frog.

	Δ AIC									
Model Name	A1	A2	A3	A5	A6	B3	B4	B5	B6	
2nd Polynomial	0.96	15.48	29.71	0.00	0.44	0.00	4.74	0.02	7.04	
3rd Order Polynomial	2.78	6.54	16.20	NA	NA	1.94	NA	1.91	6.79	
Analytis K I	1.94	15.06	14.25	5.73	0.38	0.00	0.00	0.23	4.81	
Ashrafi I	1.04	15.66	28.92	NA	0.41	0.00	NA	0.00	6.78	
Ashrafi II	1.02	15.52	29.18	NA	0.43	0.00	4.69	0.01	6.83	
Briere I	1.67	15.56	14.23	6.37	1.24	0.00	4.14	0.25	5.75	
Briere II	2.64	17.61	34.76	8.64	5.91	2.01	6.70	2.28	NA	
Eubank	3.14	12.13	32.89	3.82	2.55	0.55	4.78	0.78	10.63	
Gaussian	2.41	NA	21.00	NA	0.24	4.12	NA	0.14	4.83	
Gaussian Gomp	7.55	NA	NA	NA	4.27	4.73	11.44	5.36	NA	
Janisch I	2.88	11.50	28.83	5.28	0.00	0.53	4.03	1.03	8.41	
Janisch II	2.12	11.00	25.77	6.36	NA	1.97	3.32	1.75	0.61	
Kamykowski	2.74	NA	NA	NA	4.39	4.04	7.84	4.42	9.71	
Lactin I	2.78	13.30	32.11	4.05	0.63	0.04	3.52	0.10	10.43	
Lactin II	3.56	17.28	0.00	3.22	2.07	2.00	5.80	2.00	7.52	
Logan I	NA	17.65	NA	NA	3.95	2.00	7.47	2.14	NA	
Mitchell A	1.99	NA	19.65	NA	NA	0.62	NA	1.51	3.77	
Modified Deutsch	4.25	NA	NA	NA	NA	NA	NA	5.63	0.00	
Modified Gauss	0.00	14.86	32.81	9.24	2.21	2.02	NA	2.08	5.58	
Ratkowsky	3.81	16.16	17.05	7.73	2.41	2.01	NA	2.23	5.93	
Ruiz	4.41	0.00	23.02	9.70	1.26	2.03	3.49	NA	6.83	
Simplified Beta	2.53	11.99	29.23	4.73	0.45	0.01	6.10	0.21	8.97	
Simplified Briere	1.19	16.29	25.47	7.56	0.68	0.05	4.65	0.43	6.21	
Skew Normal	1.95	14.48	22.51	6.21	NA	2.02	3.42	2.11	2.78	
Taylor Sexton	2.10	14.78	16.96	6.31	0.34	0.51	2.57	2.12	5.64	
Thomas 2012	2.57	13.46	7.52	6.20	2.19	2.00	NA	2.12	7.23	
Tomlinson P	NA	16.95	NA	4.05	0.48	0.03	6.50	0.10	17.47	
Weibull	1.34	12.95	19.89	9.82	2.00	2.49	4.10	2.94	5.75	

2011
2012

2013 **Table 7.7** Delta AIC values for individually fit hydro-performance curves at 24 °C. Model name
2014 indicates all models fit for each frog. NAs in column values indicates that the specified model did
2015 not fit for the corresponding frog. Models with lower delta AIC were given higher weights towards
2016 the final curves for each frog.

Δ AIC									
Model Name	A1	A2	A3	A5	A6	B3	B4	B5	B6
2nd Polynomial	23.36	274.03	6.60	8.03	13.88	0.00	9.13	7.68	43.78
3rd Order Polynomial	4.00	NA	3.81	NA	NA	1.06	NA	NA	0.00
Analytis K I	30.86	251.50	2.91	4.02	0.00	0.32	NA	6.98	56.01
Ashrafi I	24.50	NA	6.09	7.97	5.50	0.01	8.88	7.65	19.86
Ashrafi II	24.01	273.82	6.27	NA	13.88	0.03	8.97	7.60	32.49
Briere I	NA	283.79	9.09	8.39	189.85	0.56	NA	6.62	81.49
Briere II	25.31	277.58	11.24	5.91	NA	1.93	12.54	9.83	62.14
Eubank	31.54	NA	2.81	8.23	NA	0.62	NA	5.82	77.13
Gaussian	29.56	269.29	4.26	8.23	0.00	0.45	4.40	6.93	70.81
Gaussian Gomp	NA	NA	14.35	NA	NA	5.34	NA	8.73	NA
Janisch I	30.80	NA	3.15	4.02	0.00	0.54	NA	6.30	74.92
Janisch II	NA	NA	3.15	5.56	NA	0.51	NA	6.81	65.54
Kamykowski	13.63	NA	0.95	10.20	NA	3.36	NA	9.29	12.61
Lactin I	NA	NA	0.22	8.06	25.49	0.50	NA	5.26	76.34
Lactin II	31.34	NA	0.39	9.77	NA	2.14	8.87	8.01	63.72
Logan I	NA	NA	NA	NA	NA	4.54	NA	12.35	NA
Mitchell A	28.12	235.10	5.00	3.54	0.00	0.36	NA	7.23	66.33
Modified Deutsch	NA	271.86	0.82	8.83	NA	NA	0.45	7.48	68.88
Modified Gauss	24.01	NA	2.05	9.68	NA	NA	NA	9.23	63.79
Ratkowsky	29.63	9.25	1.68	10.05	NA	2.28	NA	7.72	65.98
Ruiz	31.92	272.96	5.64	NA	NA	2.45	6.42	0.00	72.82
Simplified Beta	NA	NA	0.00	8.05	0.00	0.46	8.46	6.15	74.64
Simplified Briere	29.49	272.09	3.62	8.07	6.33	0.06	8.04	7.34	54.00
Skew Normal	12.69	0.00	NA	6.03	212.96	NA	0.00	7.10	64.96
Taylor S	5.50	263.63	4.94	0.00	NA	0.37	NA	8.86	69.91
Thomas 2012	10.46	6.64	1.60	9.78	NA	1.86	NA	7.27	7.41
Tomlinson P	NA	NA	NA	8.06	NA	1.04	NA	NA	NA
Weibull	0.00	6.18	2.95	5.71	213.61	0.34	12.82	7.93	41.77

2017

2018 **Table 7.8** Delta AIC values for individually fit hydro-performance curves at 15 °C. Model name
2019 indicates all models fit for each frog. NAs in column values indicates that the specified model did
2020 not fit for the corresponding frog. Models with lower delta AIC were given higher weights towards
2021 the final curves for each frog.

	Δ AIC									
Model Name	A1	A2	A3	A5	A6	B3	B4	B5	B6	
2nd Polynomial	2.07	0.00	0.29	NA	8.18	1.51	4.49	0.00	2.71	
3rd Order polynomial	NA	0.29	NA	NA	6.91	NA	6.10	NA	2.29	
Analytis K I	0.76	4.31	0.00	0.00	5.03	NA	6.83	0.84	3.68	
Ashrafi I	3.47	19.41	3.62	5.65	NA	1.54	14.95	0.10	NA	
Ashrafi II	2.42	17.05	0.50	0.07	NA	1.53	8.98	0.09	4.60	
Briere I	2.92	18.31	3.68	NA	16.67	1.62	12.60	1.42	4.39	
Briere II	4.34	10.11	1.81	2.38	22.05	3.62	6.21	3.46	4.61	
Eubank	0.16	12.03	4.23	13.46	14.43	1.04	10.73	0.93	0.00	
Gaussian	0.13	6.64	2.93	NA	12.02	2.81	7.26	0.84	2.34	
Gaussian Gomp	4.73	NA	9.24	15.44	NA	3.60	17.12	NA	6.67	
Janisch I	0.00	10.10	3.70	11.55	13.60	1.17	9.44	0.93	1.04	
Janisch II	1.61	4.80	5.57	10.86	13.19	1.18	2.18	2.84	0.62	
Kamykowski	5.04	2.59	2.64	NA	NA	5.47	8.24	4.85	5.79	
Lactin I	0.57	13.52	3.65	15.79	17.49	0.65	10.83	0.84	4.57	
Lactin II	2.91	3.36	3.13	NA	15.50	3.52	7.99	2.26	5.59	
Logan I	5.15	NA	17.39	NA	NA	NA	NA	NA	8.91	
Mitchell A	NA	2.35	2.26	6.81	10.89	NA	5.90	0.93	NA	
Modified Deutsch	2.00	2.88	5.18	NA	7.22	NA	0.00	NA	3.45	
Modified Gauss	3.30	NA	NA	NA	0.00	3.36	3.00	2.81	5.26	
Ratkowsky	2.51	2.54	3.91	8.01	12.45	3.31	7.43	2.84	4.73	
Ruiz	1.40	9.60	4.97	12.93	14.02	3.64	9.27	1.91	4.61	
Simplified Beta	0.44	12.48	4.01	13.56	16.21	0.00	9.69	0.87	4.30	
Simplified Briere	1.73	1.61	0.66	3.94	13.54	1.69	5.50	0.85	3.70	
Skew Normal	2.11	1.42	4.27	11.22	11.51	2.19	2.08	2.84	2.17	
Taylor S	0.59	8.27	2.45	5.53	5.45	NA	4.18	0.90	1.66	
Thomas 2012	2.50	0.51	1.58	NA	6.28	NA	5.49	2.90	3.80	
Tomlinson P	3.60	19.60	4.41	14.47	NA	1.62	15.21	0.85	NA	
Weibull	3.89	15.50	0.69	12.81	8.48	3.22	6.13	2.93	3.34	

2022

2023

Table 7.9 Delta AIC values for individually fit hydro-performance curves at 8 °C. Model name indicates all models fit for each frog. NAs in column values indicates that the specified model did not fit for the corresponding frog. Models with lower delta AIC were given higher weights towards the final curves for each frog.

	Δ AIC								
Model Name	A1	A2	A3	A5	A6	B3	B4	B5	B6
2nd Polynomial	8.83	14.29	0.05	15.50	4.75	8.90	18.73	13.15	0.19
3rd Order Polynomial	NA	0.00	NA	5.62	NA	2.00	13.04	10.57	1.83
Analytis K I	1.92	15.51	0.00	4.42	3.11	0.00	14.11	10.59	0.00
Ashrafi I	NA	17.78	0.69	NA	NA	8.84	NA	NA	3.16
Ashrafi II	NA	14.45	0.07	NA	8.47	8.91	NA	NA	1.39
Briere I	NA	15.71	0.38	NA	NA	9.71	NA	15.79	2.97
Briere II	5.38	15.27	2.00	NA	5.32	11.73	NA	16.62	2.06
Eubank	15.25	16.71	1.17	11.06	6.94	9.18	10.05	14.91	1.65
Gaussian	13.63	15.73	0.54	10.20	6.25	10.98	15.25	14.20	1.16
Gaussian Gomp	NA	19.90	2.59	NA	10.84	11.37	NA	NA	4.91
Janisch I	14.66	16.36	0.98	9.71	6.79	9.60	12.46	14.65	1.86
Janisch II	NA	4.63	1.32	11.32	NA	11.37	7.39	NA	NA
Kamykowski	0.00	12.96	3.84	NA	14.56	13.85	11.05	8.36	4.06
Lactin I	17.48	16.68	0.73	14.47	NA	9.99	4.18	15.84	3.45
Lactin II	14.96	17.25	2.24	0.00	8.59	9.88	0.00	16.54	2.63
Logan I	NA	NA	2.40	NA	13.90	NA	NA	18.84	8.39
Mitchell A	12.50	15.29	0.51	11.74	NA	0.77	16.58	13.90	NA
Modified Deutsch	NA	14.70	2.08	5.47	NA	11.74	NA	NA	NA
Ruiz	15.71	17.73	2.54	11.88	8.34	11.82	13.89	16.20	3.24
Simplified Beta	16.85	16.16	0.56	9.94	7.79	9.40	5.80	15.57	2.28
Simplified Briere	13.17	14.87	0.11	6.48	6.16	9.85	15.79	14.77	0.67
Skew Normal	NA	11.79	2.01	8.76	NA	11.27	4.54	0.00	NA
Taylor Sexton	15.24	17.83	0.76	2.27	2.54	9.55	12.35	9.52	1.49
Thomas 2012	0.31	15.01	1.94	1.73	4.77	11.23	6.15	12.22	1.28
Tomlinson P	NA	17.85	0.78	NA	NA	9.21	NA	NA	3.38
Weibull	18.76	9.90	2.18	NA	0.00	11.61	NA	8.18	NA

8 APPENDIX B

8.1 Mathematical Models Tested

To describe the effect of temperature (°C) and dehydration on performance, I used a candidate set of 31 non-linear models to generate performance curves (Table 8.1).

Table 8.1 Functions for all models used to fit the final thermal and hydric performance curves. $P(z)$ indicates performance, and z is the parameter describing the fixed effect of either water loss or temperature for all models. Parameter definitions are described in the parameter column.

Model Name	Fixed Response Variable (z)	Model Function	Parameters
2nd Order Polynomial (2024)	Water loss or Temperature	$P(z) = az^2 + bz + c$	a: shape parameter that defines the rate at 0 b and c: shape parameters with no biological meaning.
3rd Order Polynomial (2024)	Water loss or Temperature	$P(z) = az^3 + bz^2 + cz + d$	a, b, c, and d: shape parameters with no biological meaning.
Analytis Kontodimas I (2004)	Water loss or Temperature	$P(z) = a(z - T_{min})^2(T_{max} - z)$	a: scale parameter defining curve height T_{min}: low temperature (T) or hydration (H) at which the rate becomes negative T_{max}: high T or H at which rate becomes negative

Ashrafi I (2018)	Water loss or Temperature	$P(z) = a + b(z)^2 * \log(z) + cz^3$	a, b and c: shape parameters with no biological meaning.
Ashrafi II (2018)	Water loss or Temperature	$P(z) = a + b * d(z)^{\frac{3}{2}} + d * z^2$	a, b, c, and d: shape parameters with no biological meaning.
Biere I (1999)	Water loss or Temperature	$P(z) = a * z * (z - T_{min}) * \sqrt{(T_{max} - z)}$	a: scale parameter defining maximum height of the curve T_{min}: as above T_{max}: as above z: temperature or hydration - fixed
Briere II (1999)	Water loss or Temperature	$P(z) = a * z * (z - T_{min}) * (T_{max} - z)^{\frac{1}{m}}$	a: as in <i>Briere I</i> T_{min}: as above T_{max}: as above m: shape parameter to adjust curve asymmetry
Biere I (added intercept)	Water loss or Temperature	$P(z) = a * z * (z - T_{min}) * \sqrt{(T_{max} - z)} + d$	a: as in <i>Briere I</i> T_{min}: as above T_{max}: as above m and d: shape parameters to adjust curve asymmetry and starting point.
Eubank (1973)	Water loss or Temperature	$P(z) = a \left((z - T_{pk})^2 + b \right)$	a: scale parameter defining curve height. T_{pk}: optimum T or H b: shape parameter

Gaussian (2006)	Water loss or Temperature	$P(z) = P_{opt} * \exp\left(-0.5 * \left(\frac{abs(z - T_{pk})}{a}\right)^2\right)$	<i>a</i> : defines curve width <i>T_{pk}</i> : optimum T or H <i>P_{opt}</i> : maximum performance at <i>T_{pk}</i>
Gaussian-Gompertz (2006)	Water loss or Temperature	$P(z) = d * \exp\left(-\exp(b * (z - T_{pk}) - \theta) - a * (z - T_{pk})^2\right)$	<i>m</i> : scale parameter defining curve height <i>a</i> : shape parameter <i>T_{pk}</i> : optimum T or H
Janisch I (1925)	Water loss or Temperature	$P(z) = \left(\frac{1}{\left(\left(\frac{m}{2}\right) * (a^{z-T_{pk}} + a^{-(z-T_{pk})})\right)}\right)$	<i>m</i> : scale parameter defining curve height <i>a</i> : shape parameter <i>T_{pk}</i> : optimum T or H
Janisch II (1925)	Water loss or Temperature	$P(z) = \left(\frac{1}{\left(\left(\frac{m}{2}\right) * (a^{z-T_{pk}} + b^{-(z-T_{pk})})\right)}\right)$	<i>m</i> : scale parameter defining curve height <i>a</i> : shape parameter for curve incline <i>b</i> : shape parameter for curve decline <i>T_{pk}</i> : optimum T or H
Kamykowski (1985)	Temperature	$P(z) = a * (1 - \exp(-b * (z - T_{min}))) * (1 - \exp^{-c * (T_{max} - z)})$	<i>a</i> , <i>b</i> and <i>c</i> : shape parameters with no biological meaning
Lactin I (1995)	Water loss or Temperature	$P(z) = \exp(\rho * z) - \exp\left(\rho * T_{max} - \left(\frac{(T_{max} - z)}{DT}\right)\right)$	<i>ρ</i> : shape constant determining steepness of curve incline <i>T_{max}</i> : T or H at which the curve begins to decelerate (maximum) <i>DT</i> : thermal or hydric safety margin

Lactin II (1995)	Water loss or Temperature	$P(z) = \exp(\rho * z) - \exp\left(\rho * T_{\max} - \left(\frac{(T_{\max} - z)}{DT}\right)\right) + \lambda$	<p>ρ: shape constant determining steepness of curve incline λ: shape constant for curve height T_{\max}: T or H at which the curve begins to decelerate (maximum) DT: thermal or hydric safety margin</p>
Logan I (1976)	Water loss	$P(z) = \psi * \left(\exp(\rho * z) - \exp(\rho * T_{\max} - \left(\frac{(T_{\max} - z)}{DT}\right)) \right)$	<p>ρ: shape constant determining rate of curve incline ψ: shape constant determining rate of incline after lower threshold T_{\max}: T or H at which the curve begins to decelerate (maximum) DT: thermal or hydric safety margin</p>
Mitchell-Angilletta (2009)	Water loss or Temperature	$P(z) = \left(\frac{1}{2 * b}\right) * \left(1 + \cos\left(\left(\frac{(z - T_{pk})}{b}\right) * \pi\right)\right) * a$	<p>T_{pk}: optimum T or H a: scale parameter b: scale parameter for performance breadth</p>
Modified Deutsch (2008; 2011)	Water loss	$P(z) =$ $\begin{aligned} & \text{if } z \leq T_{pk}: P_{opt} * \exp\left(-\left(\frac{z - T_{pk}}{2\sigma}\right)^2\right) \\ & \text{if } z > T_{pk}: P_{opt} - P_{opt} * \left[1 - \left(\frac{z - T_{pk}}{T_{pk} - T_{\max}}\right)^2\right] \end{aligned}$	<p>P_{opt}: maximum performance at T_{pk} T_{pk}: optimum T or H σ: shape parameter defining steepness of curve decline T_{\max}: T or H at which the curve begins to decelerate (maximum)</p>

Modified Gaussian (2006)	Water loss	$P(z) = P_{opt} * \exp\left(-0.5 * \left(\frac{abs(z - T_{pk})}{a}\right)^b\right)$	<p>T_{pk}: optimum T or H; a: shape parameter for curve width b: scale parameter for curve asymmetry P_{opt}: maximum performance at T_{pk}</p>
Ratkowsky (modified) (1983)	Water loss or Temperature	$P(z) = \left(a * (z - T_{min}) * (1 - \exp(b * (z - T_{max})))\right)^2$	<p>a: defined as $\sqrt{P(z)/(z - T_{min})}$ T_{min}: low T or H at which rate becomes negative T_{max}: high T or H at which rate becomes negative b: parameter for fitting data beyond optimum</p>
Ruiz (2019)	Water loss or Temperature	$P(z) = B_0 + DB_{max} * \exp(-a * (z - T_{pk})^2)$	<p>B_0: parameter describing shifts in rate (or minimum performance level) DB_{max}: parameter controlling the height of the curve above baseline B_0. a: scale parameter defining curve decline rate after T_{pk}. T_{pk}: optimum T or H</p>

Rezende-Bosinovic (2019)	Temperature	$P(z) =$ $\text{if } z \leq T_{th}: B_0 * \exp\left(\frac{\log(Q_{\{10\}})}{\frac{10}{z}}\right)$ $\text{if } z > T_{th}: B_0 * \exp\left(\frac{\log(Q_{\{10\}})}{\frac{10}{z}}\right) * (1 - d * (b - z)^2)$	<p>Q₁₀: the fold change in performance as a result of increasing the temperature by 10 °C</p> <p>B₀: parameter describing shifts in rate (or minimum performance level)</p> <p>b: parameter threshold T or H beyond which the curve declines</p> <p>d: parameter controlling the rate of decline beyond the threshold temperature, T_{th}</p>
Simplified Beta (2008)	Water loss or Temperature	$P(z) = \left(rho * \left(alpha - \left(\frac{z}{10} \right) \right) \right) * \left(\frac{z}{10} \right)^{beta}$	<p>rho, beta, and alpha: shape parameters with no biological meaning.</p>
Simplified Briere (1999)	Water loss or Temperature	$P(z) = \left(a * (z - T_{min}) * \sqrt{(T_{max} - z)} \right)$	<p>a: scale parameter for maximum rate of curve</p> <p>T_{min}: low T or H at which rate becomes negative</p> <p>T_{max}: high T or H at which rate becomes negative</p>
Simplified Briere II (1999)	Water loss or Temperature	$P(z) = \left(a * (z - T_{min}) * (T_{max} - z)^{\frac{1}{m}} \right)$	<p>a: scale parameter for maximum rate of curve</p> <p>T_{min}: low T or H at which rate becomes negative</p> <p>T_{max}: high T or H at which rate becomes negative</p>

			<p><i>m</i>: shape parameter to adjust curve asymmetry</p>
			<p><i>a</i>: scale constant determining maximum height of the curve</p> <p><i>b</i>: scale parameter controlling the T or H at which performance is optimized</p> <p><i>sigma</i>: parameter controlling curve breadth (width)</p> <p><i>λ</i>: parameter controlling curve asymmetry</p>
Skew Normal (2012)	Water loss or Temperature	$P(z) = \left(a * \left(\exp \left(\frac{(-(z - b)^2)}{(sigma^2)} \right) \right) * \left(1 + \operatorname{erf} \left(\frac{(-\lambda * (z - b))}{sigma} \right) \right) \right)$	
Taylor Sexton (1972)	Water loss or Temperature	$P(z) = P_{opt} * \frac{[-(z - T_{min})^4 + 2(z - T_{min})^2 * (T_{pk} - T_{min})^2]}{(T_{pk} - T_{min})^4}$	<p><i>a</i>: defines curve width</p> <p>T_{min}: low T or H at which rates become negative</p> <p>T_{pk}: optimum T or H</p> <p>P_{opt}: maximum performance at T_{pk}</p>
Thomas (2012)	Water loss or Temperature	$P(z) = a * \exp^{b*z} \left(1 - \left(\left(\frac{z - T_{opt}}{\frac{c}{2}} \right) \right)^2 \right)$	<p><i>a</i>: shape constant with no biological meaning</p> <p><i>b</i>: shape constant with no biological meaning</p> <p>T_{opt}: location of the maximum height of the curve</p> <p><i>c</i>: range of T or H over which the performance rate is positive</p>

**Tomlinson-
Phillips
(2015)**

Water loss or
Temperature

$$P(z) = B_0 * (\exp(a * z) - \exp(z - b))$$

B_0 : scale parameter for performance rate at T_{min}
 a : shape constant indicating upwards slope of the curve
 b : parameter controlling peak height of the curve.

**Weibull
(2024; 2010)**

Water loss or
Temperature

$$P(z) = P_{opt} * \left(\left(\frac{c-1}{c} \right)^{\frac{1-c}{c}} \right) * \left(\left(\left(\frac{z - T_{pk}}{d} \right) + \left(\frac{c-1}{c} \right)^{\frac{1}{c}} \right)^{(c-1)} \right) \\ * \exp \left(- \left(\left(\frac{z - T_{pk}}{d} \right) + \left(\frac{c-1}{c} \right)^{\frac{1}{c}} \right)^c + \left(\frac{c-1}{c} \right) \right)$$

P_{opt} : maximum performance value
 T_{pk} : T or H at which performance is optimized
 d : parameter describing the curve breadth
 c : parameter defining the overall curve shape