# Warming in the land of the midnight sun: breeding birds may suffer greater heat stress at high- vs low-Arctic sites

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#### 1 Abstract

2 Rising global temperatures are expected to increase reproductive costs for wildlife as greater 3 thermoregulatory demands interfere with essential breeding activities such as parental care. 4 However, predicting the temperature threshold where reproductive performance is negatively 5 impacted remains a significant hurdle. Using a novel thermoregulatory polygon approach, we 6 predicted the threshold temperature at which an Arctic songbird-the snow bunting 7 (*Plectrophenax nivalis*)—would need to reduce activity and perform below the 4-times basal 8 metabolic rate (BMR) required to sustain nestling provisioning to avoid overheating. We then 9 compared this threshold to operative temperatures recorded at high  $(82^{\circ}N)$  and low  $(64^{\circ}N)$ 10 Arctic sites to estimate how heat constraints translate into site-specific impacts on sustained 11 activity level. We predict buntings would become behaviourally constrained at operative 12 temperatures above 11.7°C, whereupon they must reduce provisioning rates to maintain thermal 13 balance. Low Arctic sites had larger fluctuations in solar radiation, producing consistent daily 14 periods when operative temperatures exceeded 11.7°C. However, high-latitude birds faced entire, 15 consecutive days where parents would not be able to sustain required provisioning rates. These 16 data indicate that Arctic warming is likely already disrupting the breeding performance of cold-17 specialist birds, but also suggests counterintuitive and severe negative impacts of warming at 18 high-latitude breeding locations.

#### 20 **1. Introduction**

21 Animals frequently experience stages that demand significant increases in their sustained rate of 22 energy expenditure (e.g., reproduction [1-3]). In an era of rapid climate change that is impacting 23 species and ecosystems worldwide [4], understanding energetic limits and their causes is 24 paramount for predicting whether organisms can respond to current rising global temperatures 25 [5]. Historically, energetic limits among endotherms have either been attributed to intrinsic 26 physiological factors (e.g., the digestive capacity of the gut to assimilate energy; central 27 limitation hypothesis [6]) or constraints residing in the metabolic capacity of specific peripheral 28 tissues (e.g., mammary glands or muscle tissue; peripheral limitation hypothesis [1]). Recently, 29 Speakman and Król [7–9] proposed an alternative hypothesis, termed the heat dissipation limit 30 (HDL) theory, which contends that the maximal rate of energy expenditure for an endothermic 31 animal is limited by physiological factors governing heat dissipation capacity and the consequent avoidance of lethal body temperatures. Importantly, whereas the peripheral limitation hypothesis 32 33 argues that energetic constraints may act on a range of tissues and organs (e.g., mammary glands, 34 brown adipose tissue, or skeletal muscle), the HDL theory proposes a universal constraint in the 35 form of heat dissipation and thereby provides a mechanistic link between an animal's 36 physiological capacity to maximize energy expenditure with the interplay between heat 37 dissipation and ambient temperature.

38 Despite the significant conceptual gains that the HDL theory has provided in linking heat 39 dissipation capacity with energetic expenditure, our ability to predict the ambient temperatures 40 that will ultimately constrain an animal's performance (i.e., sustained rate of energy expenditure) 41 remains a major impediment to assessing species vulnerability to climate change [10]. Although 42 several studies have reported threshold temperatures above which sustained activity and/or 43 reproductive performance were compromised [10–14], these studies derived threshold values 44 from post-hoc analyses on behavioral observations (e.g., provisioning rates) and are therefore not predictive by design. Recently, Rezende and Bacigalupe [15] proposed a predictive analytical 45 46 tool - the thermoregulatory polygon - for estimating the dimensional space in which 47 thermoregulation is possible given an animal's combined rate of energy expenditure and the 48 environmental temperatures it is operating within. Thermoregulatory polygons are built from 49 commonly measured physiological variables (basal and maximal metabolic rate, and minimum 50 and maximum thermal conductance) to delineate the boundaries in which heat production and

dissipation are balanced [15]. Thus, thermoregulatory polygons can help estimate animal
responses to further warming by integrating concepts of the HDL theory to predict the ambient
temperatures over which endothermic animals can sustain activity and avoid lethal body
temperatures. Surprisingly, despite their potential as a predictive tool, to our knowledge only one
study has applied thermoregulatory polygons, using them to predict the energetic consequences
of activity time in nocturnal and diurnal mammals [5].

57 Among endotherms, birds are expected to be particularly sensitive to increasing 58 environmental temperatures [16,17]. The offspring-rearing period for parents with dependent 59 young requires substantial increases in sustained work effort, with adults typically performing at 60 4 to 6 times their basal metabolic rate (i.e., resting rate of energy expenditure; [2,3,6]). Any 61 excess heat generated as a by-product from foraging and provisioning must ultimately be 62 dissipated, or birds risk overheating (hyperthermia). Problematically, birds often breed during the 63 warmest parts of the year when it is hardest to passively shed body heat [18]. Indeed, birds often 64 decrease activity on days with warmer ambient temperatures, which is likely a thermoregulatory 65 response to avoid heat stress [19–23]. Recent studies have also shown that when a bird's capacity 66 to dissipate body heat is increased (e.g., by experimentally removing insulative feathers), 67 provisioning adults can sustain higher levels of activity and invest more in both their current and 68 future reproductive efforts [24–27]. Thus, reproductive performance can be constrained by a 69 bird's capacity to dissipate body heat produced during essential breeding activities, suggesting 70 that increasing environmental temperatures could significantly impact reproductive investment. 71 Here, we apply a thermoregulatory polygon to snow buntings (*Plectrophenax nivalis*; figure 72 1b), an Arctic-breeding songbird, to investigate how environmental temperature affects the 73 interaction between thermoregulation and sustained energy expenditure on the breeding grounds. 74 Applying thermoregulatory polygons to Arctic endotherms is extremely pertinent and valuable 75 for predicting how increasing ambient temperatures under climate change will impact essential 76 life-history stages through thermal constraints on behavior. Many Arctic animals are cold 77 specialists and regularly endure extremely cold weather and have evolved physiological 78 adaptations for minimizing heat loss [28–31]. Consequently, high-latitude breeding species are 79 likely vulnerable to moderate increases in ambient temperature [32–35]; an alarming fact given 80 that the Arctic has warmed faster than the global average and is expected to continue outpacing 81 the global average over the 21<sup>st</sup> century [4]. In addition, O'Connor et al. [32] recently showed

that buntings in particular can become heat-stressed at even moderate air temperatures and that their evaporative cooling capacity is extremely limited. Consequently, highly active, breeding snow buntings exposed to constant solar radiation and modest rises in air temperature would be largely dependent on behavioural thermoregulatory strategies (e.g., reducing provisioning effort) as opposed to physiological mechanisms (e.g., sustained increases in evaporate water loss rates) to dissipate body heat and avoid lethal body temperatures.

88 Our goal was to estimate how sensitive snow buntings' performance may be to increasing 89 Arctic temperatures, given their heat dissipation capacity. We first used thermal physiological 90 data to construct a thermoregulatory polygon and predict the threshold temperatures at which 91 sustainable performance (e.g., birds actively provisioning nestlings) would be expected to 92 decline in buntings maintaining thermal balance (i.e., heat produced = heat dissipated). We then compared the thermoregulatory polygon prediction to both operative and air temperatures 93 94 measured in the field at two latitudinally distant breeding sites to evaluate how heat constraints 95 on bunting performance (i) differed between a low and high Arctic region, and (ii) could 96 translate into site-specific impacts on reproductive performance and success.

97

### 98 2. Materials and methods

## 99 (a) Operative and air temperature measurements

100 We measured operative and air temperatures during the snow bunting breeding period at two 101 research sites in northern Canada representing the low-Arctic (East Bay Island; 64°01'N, 102 81°47'W; [36]) and high-Arctic (Alert; 82°30'N, 62°20'W; [37]; figure 1*a*). Operative 103 temperature represents the temperature of the thermal environment as perceived by an individual 104 and integrates the physical properties of the animal with the thermal properties of the local 105 environment (e.g., air temperature, radiation, and wind; [38]). To measure the operative 106 temperature perceived by snow buntings at our two sites, we used 3D printed, hollow plastic 107 model birds (hereafter 3D models; [39,40]; figure 1c). We printed the 3D models to match the 108 size and shape of an adult snow bunting (see electronic supplementary material, figure S1 in 109 appendix S1). Additionally, we painted the 3D models to match the spectral properties of male 110 snow buntings in breeding plumage, thereby allowing the 3D models to act as operative 111 temperature thermometers. We painted the 3D models to match the color morph of male

112 buntings because males feed females while incubating [41] and consequently are exposed to 113 solar radiation for a longer duration throughout the breeding period. We also focused on males 114 given their simplified monochromatic breeding plumage [42] and because males actively 115 provision offspring at similar rates to females [43]. We used a spectrophotometer (Ocean Optics 116 Jaz spectrometer) to measure the spectra of the black (N = 16 birds) and white (N = 27 birds) 117 feather regions of male snow buntings within the 300-700nm wavelengths. We used the *pavo* 118 package in R [44] to convert the spectra wavelengths to a red:green:blue (R:G:B) color 119 combination. We then used an R:G:B-to-paint converter (https://www.e-paint.co.uk/convert-120 rgb.asp) to acquire a paint that best matched the R:G:B color combination of male bunting 121 feathers. We opted to paint the 3D models (N = 68 at the high-Arctic site and N = 13 at the low-122 Arctic site) instead of placing the skin and plumage of a male snow bunting over the models as 123 this optimized our experimental design by allowing us to record operative temperature in numerous models simultaneously across a broader geographic area [45]. 124

125 We measured the internal temperature of each 3D model by placing a temperature logger in the centre of each model. At the high-Arctic site, we drilled a hole in the belly of each 3D model 126 127 and secured an iButton (model DS1921G-F5, Maxim Integrated, San Jose, CA USA) in the 128 approximate center (figures S2 and S3 in appendix S1) by gluing it to the end of a wooden dowel 129 surrounded by a rubber stopper, creating an airtight seal around the drill-hole (figure S4 in 130 appendix S1). At the low-Arctic site, models were similarly set up except for using Hobo data 131 loggers (Pendant model, MX2201, Onset Inc., Bourne, MA USA) instead of iButtons, which we 132 secured with silicone caulking. At both sites, the 3D printed models were secured to a wooden 133 plank by gluing a wooden dowel to a notch in the 3D model (figures S3 and S4 in appendix S1). 134 We cut the wooden dowels to approximate the height of a standing snow bunting. We covered 135 each plank in the field using the substrate beneath the models to mimic the thermal properties of 136 snow bunting's natural environment (e.g., snow, moss, or rocky shale; figure S5 in appendix S1). 137 At each site, we deployed 3D models within representative breeding territories and across 138 naturally occurring habitats to adequately capture the thermal heterogeneity experienced by 139 buntings. In the high-Arctic, we recorded operative temperatures every 5 minutes from 22 May 140 to 7 September 2019 and models were deployed over six separate periods, each lasting 141 approximately 7 days (due to iButton memory limitations). After 7 days, we downloaded the 142 operative temperature data and redeployed the 3D models to a new location. In the low-Arctic,

we originally deployed 22 models once and recorded operative temperatures continuously from
11 June to 19 July 2019 at 2-min intervals. Unfortunately, polar bears damaged many models, so
only 13 of our original 22 models were usable. However, the distribution of snow bunting
breeding pairs at the low-Arctic site is much smaller in area than in the high-Arctic and these 13
models thus still provided sufficient geographic coverage of East Bay's microclimates
experienced by snow buntings.

At both study sites, we collected air temperature data to compare against operative temperatures. In the high-Arctic, the downloaded meteorological data was measured at the National Oceanic and Atmospheric Administration's (NOAA) broadband radiation station located adjacent to the Global Atmospheric Watch (GAW) Observatory (82°28'N, 62°30'W). These data are 1-min averages of air temperature obtained at a height of 3 m above the ground using an aspirated Vaisala HMP-235 (PT100 sensor). In the low-Arctic, we collected an air temperature value every 30-min using six Kestrel weather meters (model 5500, Boothwyn, PA,

156 USA) placed 2-3 m above ground level at separate locations across the study site.

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## 158 **(b)** Thermoregulatory polygon parameters and construction

159 Thermoregulatory polygons use an animals rate of energy expenditure and thermal conductance 160 properties to delineate the space in which endotherms can balance heat production with heat loss 161 under a given environmental temperature [15]. We calculated the basal metabolic rate (BMR; N 162 = 28 birds), minimum wet thermal conductance ( $C_{min}$ ; N = 20 birds) and maximum dry thermal 163 conductance ( $C_{max}$ ; N = 21 birds) on a wild population of snow buntings in the high-Arctic from 164 2 June to 25 July 2018. Information on gas analyzers, experimental protocol, body and air 165 temperature measurements, and equations used for calculating metabolic rates are described in 166 detail in Le Pogam et al. [37,46,47] and O'Connor et al. [32]. Briefly, we measured BMR 167 overnight on fasted individuals resting inside a darkened metabolic chamber at thermoneutral 168 temperatures (mean air temperature =  $26.2 \pm 0.8^{\circ}$ C; note, air temperature inside metabolic 169 chambers is equivalent to operative temperature [38]). For  $C_{min}$ , we measured metabolic rates on 170 individuals at a constant air temperature below their lower critical temperature of  $10^{\circ}C$  ([28]; 171 mean  $C_{min}$  air temperature = -19.0 ± 1.8°C). We did not measure rates of evaporative water loss 172 during our Cmin runs and therefore for each bird we calculated minimum wet thermal 173 conductance as follows:

$$C_{min} = \frac{MR}{T_b - T_a} \qquad \qquad Eq. \ 1$$

174

175where MR represents metabolic rate in Watts, and  $T_b$  and  $T_a$  are the mean body and air176temperatures, respectively. At air temperatures below the lower critical temperature, evaporative177heat loss is minimal and thus its inclusion has little influence on  $C_{min}$  [48]. During metabolic178measurements for  $C_{min}$ , we measured  $T_b$  at the start and end of each run and used the mean value179for our calculations.

180 We determined maximum dry thermal conductance by exposing birds to a ramped 181 temperature protocol of gradually increasing air temperatures [32]. We only included birds that tolerated air temperatures above 31.5°C, representing the mean air temperature minus the 182 183 standard deviation at which buntings started panting [32], as we assumed that birds that had 184 initiated panting had reached their maximum thermal conductance [49]. This resulted in the 185 removal of only 1 bird from the final data set. At higher air temperatures, evaporative heat loss 186 becomes significant and therefore must be accounted for in the calculation of maximum thermal 187 conductance [48]. We thus calculated maximum dry thermal conductance for each bird as 188 follows:

$$C_{max} = \frac{(MR - EHL)}{(T_b - T_a)} \qquad Eq. 2$$

189

190 where EHL represents evaporative heat loss measured during respirometry trials [32]. During 191  $C_{max}$  experiments, we measured T<sub>b</sub> continuously and therefore were able to calculate an average 192 T<sub>b</sub> over the same 5-min time window that metabolic rates were calculated [32].

To generate the thermoregulatory polygon, we calculated a combined mean across birds for each parameter (i.e, BMR,  $C_{min}$ ,  $C_{max}$  and  $T_b$ ). The BMR mean became the lower boundary of the thermoregulatory polygon. The  $C_{min}$  and  $C_{max}$  means became the slopes of the left and right boundaries, respectively. We calculated the y-intercepts for the  $C_{min}$  and  $C_{max}$  slopes using the equation:

$$MR = C(T_a) + b \qquad Eq. 3$$

198 where C represents the combined  $C_{min}$  or  $C_{max}$  mean across birds and b is the y-intercept. We

assumed  $T_a = T_b$  when MR = 0 [48] and used the combined  $T_b$  mean across birds during  $C_{min}$ 

200  $(41.0 \pm 0.4^{\circ}C)$  and  $C_{max} (42.6 \pm 0.7)$  measurements.

201

# 202 (c) Estimating sustainable performance in the high-Arctic (Alert) and low-Arctic (East203 Bay)

204 We conducted all analyses in R 4.0.4 [50]. Over the course of the field season in the high-Arctic, 205 we recorded a total of 843,773 individual operative temperature values from 68 3D models and a 206 total of 107,092 air temperature values. In the low-Arctic, we recorded 405,000 individual 207 operative temperature values from 13 models and a total of 10,803 air temperature values. We 208 used these raw temperature data to create a time series of operative and air temperatures for each 209 site averaged at 1-h intervals using the *timeAverage* function in the *openair* package [51]. 210 The discontinuous sampling protocol in the high-Arctic (e.g., downloading data and 211 redeploying models) resulted in 643 1-h gaps in our operative temperature time series. To 212 estimate the percentage of time on a given day that buntings would have been behaviorally 213 constrained from heat (see below) it was necessary to fill these gaps in our data. We filled the 214 operative temperature gaps by fitting an artificial neural network [52–54] with the *neuralnet* 215 package [55] to predict operative temperature based on seven radiative and meteorological 216 variables observed at the NOAA broadband radiation station. Specifically, the input layer of the 217 neural network included air temperature, wind speed (m/s), downwelling shortwave radiation 218 flux (calculated as the sum of the contributions from diffuse and direct shortwave radiation; 219  $W/m^2$ ), reflected shortwave radiation flux ( $W/m^2$ ), albedo (calculated as the ratio of the reflected 220 shortwave radiation flux to downwelling shortwave radiation flux), net longwave radiation flux 221 (calculated by subtracting the longwave radiation flux emitted by the surface from the 222 downwelling longwave radiation flux;  $W/m^2$ ), and diffuse fraction as a measure of the 223 proportional influence of the direct sun and a proxy for cloudiness (calculated as the ratio of 224 diffuse shortwave flux to total downwelling shortwave flux). Before training and testing the 225 neural network, we applied a ranging standardization to the data, resulting in the data ranging 226 between 0 and 1 [56]. We fitted the network with one hidden layer comprised of five neurons 227 and trained the model using a random sample of 90% of the data set (1,752 values). The model 228 was tested on a random sample of 10% of the data set (195 observations). We cross-validated the 229 neural network by repeating the process (i.e., training, testing, and calculating mean square 230 prediction error) 20 times consecutively. The neural network predicted hourly operative temperatures with an average mean square error of  $1.8^{\circ}$ C (range = 1.2 to  $2.7^{\circ}$ C). 231

232 We next used the  $C_{max}$  slope (i.e., the right-side upper boundary of the thermoregulatory 233 polygon) to estimate the maximum sustainable energy expenditure of snow buntings maintaining 234 thermal balance (i.e., heat production = heat loss) under both air temperature and operative 235 temperature recordings. As the provisioning period is one of the most energetically expensive 236 life-history stages for birds [3], we placed particular emphasis on the sustainable performance 237 possible for buntings during this period. At the high-Arctic site, adult buntings are typically 238 observed provisioning young from 4 July to 25 July (A. Le Pogam, personal observations) and at 239 the low-Arctic site from 3 July to 24 July [42,43,57]. We thus used these respective date ranges 240 to represent the typical provisioning period at each site. We defined performance as a multiple of 241 BMR and assumed that 4-times BMR is the minimum sustainable performance required for adult 242 buntings to adequately provision nestlings [2,3]. Therefore, we defined 4-times BMR as the 243 energetic threshold for "optimal performance" and we calculated the percentage of time on a given day that buntings could work at either optimal ( $\geq$  4-times BMR) or suboptimal (< 4-times 244 245 BMR) performance levels based on either operative temperature or air temperature recordings. 246 Lastly, we assumed buntings rested and significantly reduced provisioning rates for 3-h a day 247 [58] and we only used temperature values measured between 01:00 and 22:00 hrs when 248 calculating the daily percentage of time that buntings could work at optimal or suboptimal 249 performance levels.

#### 250 **3. Results**

#### 251 (a) Thermoregulatory polygon

252 All values reported are mean  $\pm$  standard deviation. The mean basal metabolic rate (BMR) of 253 snow buntings was  $0.564 \pm 0.076$  W. Mean thermal conductance varied three-fold, with a 254 calculated minimum wet thermal conductance of  $0.023 \pm 0.005$  W/°C and a maximum dry 255 thermal conductance of  $0.073 \pm 0.023$  W/°C (figure 2*a*). The thermoregulatory polygon bounded 256 by these physiological parameters predicts that snow buntings can maintain thermal balance and 257 sustain optimal performance (i.e.,  $\geq$  4-times BMR) at operative temperatures of up to 11.7°C 258 (figure 2b). Once operative temperature exceeds  $11.7^{\circ}$ C, we expect buntings to become 259 behaviorally constrained by heat and, consequently, forced to perform at suboptimal levels to 260 avoid overheating. During the peak provisioning period when buntings are most active, operative

temperatures over both sites ranged from  $-0.6^{\circ}$ C to 24.6°C, leading to a 12.3°C zone in which we predict that buntings could maintain thermal balance and perform optimally (figure 2*b*).

263

264 (b) Estimated sustainable performance in the high-Arctic (Alert) and low-Arctic (East Bay)

Air and operative temperatures at the high-Arctic site increased steadily from the beginning of the breeding period until peaking during the nestling-provisioning period, and then gradually declined towards the post-fledging period (see electronic supplementary material, figure S1*a* in appendix S2). Operative temperatures experienced by buntings frequently exceeded shaded air temperature, and on average were  $3.5 \pm 3.1$  °C warmer than air temperature (range of differences between operative and air temperature = -4.9 °C to 14.5 °C; figure S1*b* in appendix S2).

271 At the high-Arctic site, only operative temperature exceeded the predicted thermoregulatory 272 polygon threshold value of 11.7°C before 5 July (figure 3a). However, from 5 July to 5 August, 273 both air temperature and operative temperature measurements periodically exceeded 11.7°C 274 (figure 3a), suggesting that buntings would have had to regularly perform at suboptimal levels 275 below 4-times BMR during this period. Within the typical nestling-provisioning period at the 276 high-latitude site (i.e., 4 July to 25 July) when buntings are most energetically active, modelling 277 based on the polygon predicts that buntings experienced multi-day periods where they could 278 have either performed at optimal levels for their entire active period (i.e., 01:00 hrs to 22:00 hrs) 279 or they would have been heat constrained and forced to work below 4-times BMR for the entire 280 active period (figure 4a). For example, based on operative temperature recordings, there were 281 two periods of consecutive days (9 to 11 July and 19 - 22 July) where we predict that buntings 282 would never have been heat constrained and could have worked at optimal performance levels 283 for their entire active period (figure 4a). However, there were two periods of consecutive days (6 284 to 8 July and 13 to 17 July) when operative temperatures exceeded 11.7°C for their entire active 285 period, and we predict that buntings would have to reduce their provisioning rates to lower 286 metabolic heat production and avoid lethal body temperatures. From 13 to 19 July the polygon 287 model also predicted that buntings experienced only 5 hours with temperatures that allowed them 288 to both maintain thermal balance and sustain an optimal performance level of  $\geq$  4-times BMR. 289 These findings suggest that temporal variation in heat constraints on the sustainable performance 290 of breeding snow buntings in the high-Arctic correspond to synoptic cycles (i.e., weather-scale, 291 2-4 days) rather than regular circadian cycles. This is explainable by the suppressed amplitude of

the diurnal cycle at the high latitude site where the sun is above the horizon continuously from early April through early September. Overall, under operative temperature recordings at the high-Arctic site, the percentage of time each day that buntings would have been behaviorally constrained from heat during their active period ranged from a minimum of 19% (4 hrs) to a maximum of 100% (21 hrs; figure 4*a*).

297 At the low-Arctic site, average hourly temperatures were relatively consistent across the 298 study period (figure S2a in appendix S2). The overall mean difference between operative and air 299 temperature was  $4.0 \pm 4.1$  (range = -2.7°C to 15.5°C; figure S2b in appendix S2). In contrast to 300 the high-Arctic site, where both operative and air temperature recordings exceeded the 301 thermoregulatory polygon threshold temperature of 11.7°C, only operative temperature 302 recordings at the low-Arctic site consistently placed a heat constraint on buntings' sustainable 303 performance (figure 3b). For example, during the typical nestling-provisioning period in the low-304 Arctic (i.e., 3 July to 24 July), modelling based on the polygon predicted that buntings would 305 have been behaviorally constrained on just four days using measured air temperature values, 306 whereas operative temperature values suggest that bunting performance would have been 307 constrained to some degree on 15 of the 17 days during which temperatures were recorded 308 (figure 4b). Furthermore, unlike the high-Arctic birds, buntings at the low-Arctic site would not 309 be forced to perform at suboptimal levels for their entire active period (i.e., an entire day), but 310 would instead be forced to alter performance for a portion of each day (figure 4b). Overall, under 311 measured operative temperatures at the low-Arctic site, the percentage of time that buntings 312 would have been behaviorally constrained from heat on a given day during their active period 313 (i.e., 01:00 to 22:00 hrs) ranged from a minimum of 5% (1 hr) to a maximum of 67% (14 hrs; 314 figure 4b).

315

## 316 **4. Discussion**

A growing body of evidence suggests that increasing environmental temperatures associated with climate change will impose reproductive costs on birds via trade-offs between essential breeding behaviours and the need to dissipate body heat and avoid lethal body temperatures [10,11,17,25,59,60]. To date, predicting the threshold temperatures that will adversely affect breeding activity has been a limiting factor in forecasting the impacts of anthropogenic climate change on birds. Additionally, studies on how thermoregulatory demands will negatively impact

breeding behaviour within birds is overwhelmingly biased towards hot, arid climates while
studies on Arctic birds are severely lacking.

325 Using a thermoregulatory polygon approach, we estimated the maximal sustained energy 326 expenditure in an Arctic songbird maintaining thermal balance across a range of environmental 327 temperatures. Assuming an optimal performance level of 4-times basal metabolic rate (BMR; 328 [3]), our findings predict that buntings will become heat constrained at operative temperatures 329 above 11.7°C. At this point, buntings would need to reduce their maximal sustained energy 330 expenditure and provision their offspring at sub-optimal performance levels to balance heat loads 331 and avoid a lethal body temperature. Additionally, by examining impacts at both a low and high 332 Arctic breeding site, our data reveal latitude-dependent operative temperature traces, likely 333 linked to differences in available sunlight and solar radiation input, culminating in site-specific 334 patterns in the heat constraints placed on an animal's maximal sustained energy expenditure. 335 Consequently, synoptic-scale influences on local temperature apparently dominate in modulating 336 operative temperatures in the high-Arctic, whereas the diurnal cycle is the dominant factor in the 337 low-Arctic. Collectively our results indicate that while Arctic warming will expose all snow 338 bunting populations to more periods above their threshold temperature for sustained optimal 339 performance, high-Arctic birds will likely face increases in the duration and magnitude of these 340 periods due to the added effects of a higher average solar radiative flux. The expectation then 341 will be that high-Arctic populations will face greater downstream costs to reproductive 342 performance (e.g., provisioning rates), and therefore ultimately breeding success compared to 343 low-Arctic populations.

344

#### 345 (a) Using the thermoregulatory polygon to predict thermal constraints

346 The heat dissipation limit theory postulates that an animal's maximum sustained energy 347 expenditure scales with its capacity to dissipate body heat [8]. Many factors influence an 348 animal's thermoregulatory ability, including basal metabolic rate and thermal conductance [61– 349 64]. Based on buntings' basal metabolic rate (0.564 W) and maximum dry thermal conductance 350  $(0.073 \text{ W})^{\circ}$ C), the thermoregulatory polygon predicts that at operative temperatures above 351 11.7°C, snow buntings cannot maintain thermal balance and sustain activity at optimal 352 expenditure rates of 4-times basal metabolic rate. Therefore, if operative temperature exceeds 353 11.7°C for extended periods, we would expect to observe (i) a decline in the body condition of

354 breeding adults as they increase evaporative water loss rates to maintain foraging effort and 355 provisioning rates outside of the thermoregulatory polygon limits [11,59,65] and/or (ii) a slower 356 growth rate, prolonged breeding period and potentially reduced fledging mass as adults reduce 357 provisioning rates to avoid lethal body temperatures [13,41,66–68]. Although not derived under 358 the thermoregulatory polygon framework, Cunningham et al. [12] reported lower provisioning 359 rates at higher ambient temperatures in common fiscals (Lanius collaris) and fledglings were 360 significantly lighter if maximum air temperature often exceeded a threshold temperature of 33°C. 361 The comparatively low threshold temperature for buntings  $(11.7^{\circ}C)$  likely stems from their 362 physiological adaptions for life in the cold [37]. Consequently, snow buntings' cold 363 specialization appears to come at the cost of not being able to adequately dissipate heat through 364 increases in maximum thermal conductance at even moderate operative temperatures. 365 Because the thermoregulatory polygon boundaries are set by the thermal conductance of the 366 animal, they represent the thermal space in which an animal can balance heat loss and gain 367 through sensible heat flow (i.e., non-evaporative mechanisms). Theoretically, an animal could 368 maintain thermal balance and sustain a high rate of energy expenditure outside its 369 thermoregulatory polygon by continuously dissipating body heat evaporatively. However, long-370 term evaporative water loss is not sustainable and evaporative cooling capacities vary 371 significantly among species [16,69,70]. Recently, O'Connor et al. [32] showed that the 372 evaporative cooling capacity of buntings is extremely limited, with most birds unable to 373 evaporatively shed an amount of heat equivalent to their metabolic heat production. Therefore, it 374 is unlikely that snow buntings can rely on evaporative cooling for prolonged periods to sustain 375 activity outside their thermoregulatory polygon limits, and, instead, will be highly dependent on 376 behavioral thermoregulation.

377

## 378 (b) Site-specific impacts of thermal constraints on breeding performance and success

Solar radiation is a major driving force of operative temperature and can vary by time of day, year, or geographic location [52,71]. Our two sites represent the general southern and northern breeding limits for Arctic-breeding snow bunting populations in Canada [58], and are separated by ~18° latitude. This latitudinal gap results in different quantities of solar radiation reaching the earth's surface [52], likely producing the significant differences observed in the duration and frequency that operative temperature exceeded the predicted threshold temperature. For example, 385 during the peak nestling-provisioning period, buntings at the high-Arctic site were predicted to 386 frequently experience consecutive days where they would not be able to perform at 4-times their 387 BMR. In contrast, buntings in the low-Arctic were predicted to experience shorter, but more 388 consistent heat constraints on provisioning activity almost every day. Given that snow bunting 389 nestlings have some of the highest recorded growth rates of any passerine (11.6-13.0 %/day of 390 adult body mass; [72]), these latitudinal differences in constraints suggest that warming will 391 produce different impacts on provisioning behaviour and hence offspring growth and survival at 392 different populations. For instance, lower-latitude breeding birds could possibly make up for 393 reduced provisioning opportunities each day by adjusting their activity budget throughout the 394 day; working harder during the cooler periods to counteract overheating risks during warmer 395 periods [5,73]. Indeed, under identical heat loads, Tapper et al. [27] observed higher feeding 396 rates in wild female tree swallows (Tachycineta bicolor) that had their ventral feathers clipped to 397 experimentally increase heat dissipation rates relative to unclipped females. Alternatively, 398 parents breeding at lower latitudes could provision growing nestlings at lower rates per day and 399 possibly extend the developmental period of the growing young. However, this could 400 nonetheless impose survival constraints on nestlings and fledglings given that ground-nesting 401 songbird species have evolved rapid growth rates and shorter in-nest development periods due to 402 high rates of nest predation [74,75], as well as the short, ephemeral nature of productivity in 403 insects required for offspring growth [76–78].

404 For higher latitude populations, the accumulation of reduced provisioning opportunities for 405 consecutive days could impose substantial developmental costs on nestling development that 406 may simply be too great for parents to compensate for on cooler days. Chick provisioning in 407 buntings typically lasts 13 days; lowering provisioning rates for 3-4 consecutive days could have 408 major impacts on chick condition at fledging, and possibly, post-fledging survival [60,79,80]. 409 Consequently, as rapid Arctic warming continues [4], the temperature-dependent costs on 410 reproductive performance may be more strongly felt at higher latitudes where climatic and 411 meteorological patterns subject individuals to unique operative temperature cycles, with above 412 threshold temperatures potentially lasting for days at the peak of breeding activities. It is worth 413 noting, however, that our 3D models were painted to match the male color morph and therefore 414 represent operative temperatures perceived by male snow buntings. During the provisioning 415 period, both male and female buntings feed young and thus the operative temperatures

416 experienced by females may differ from males leading to different sex constraints on
417 performance. For instance, females lack the full dark back of male buntings and hence may
418 experience lower operative temperatures allowing them to maintain higher provisioning rates
419 than males. Nevertheless, under such a scenario, we would still predict negative impacts on
420 nestling condition and fledgling success as both parents cannot adequately feed young at optimal
421 rates.

422 The thermal environment experienced by wild animals represents a complex integration 423 of biotic and abiotic factors operating at spatial scales relevant to the size of the animal 424 [38,81,82]. The thermal environment, as measured by operative or standard operative 425 temperature [83], can significantly deviate from air temperature [32,84]. Indeed, we found that 426 operative temperature was, on average, 3.5 to 4.0 °C above air temperature but could exceed it 427 by as much as 14.5 or 15.5°C. More importantly, our data show that the duration and frequency 428 that the threshold temperature was exceeded markedly differed depending on the heat index used 429 (i.e., air temperature or operative temperature). The implication of these differences is significant 430 because any derived predictions for maximal sustained energy expenditure under increasing 431 global temperatures will be substantially different whether air temperature or operative 432 temperature measurements are used. For example, using air temperature alone at the low-Arctic 433 site, we would predict that snow buntings are seldom heat constrained and can currently sustain 434 optimal energy expenditures, despite operative temperatures indicating otherwise. In contrast, at 435 the high-Arctic site during the peak provisioning period, air temperature measurements often 436 exceeded the temperature threshold alongside operative temperature but the percentage of time 437 that buntings were constrained was greater under operative temperature. Consequently, using air 438 temperature alone, especially values derived at macroscales (e.g., WorldClim dataset at 1 km<sup>2</sup> 439 [85]), will certainly misrepresent an animals realized thermal environment, which operates at 440 microscales, leading to biased and erroneous predictions on the impacts of climate change 441 [86,87].

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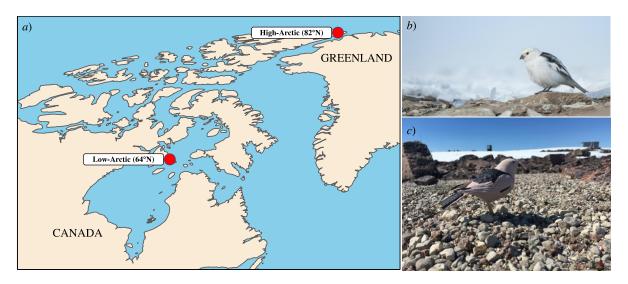
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- 682

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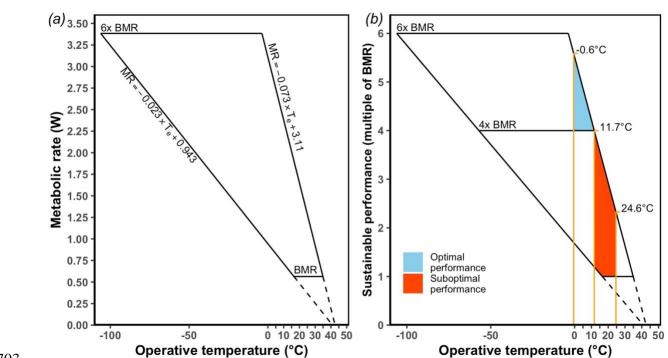
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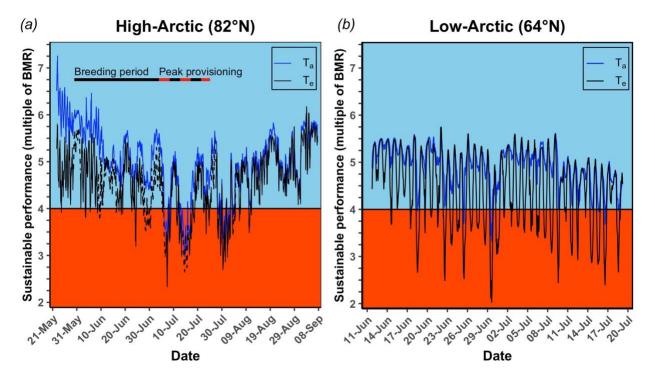
699 Figure 1. *a*) The location of the low-Arctic (East Bay Island, 64°N) and high-Arctic (Alert,

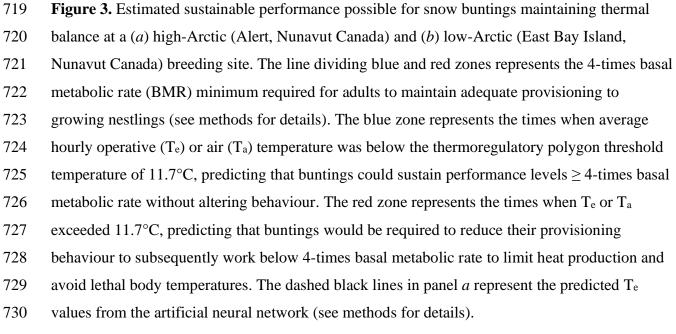
- 700 82°N) study sites. b) Photo of a snow bunting at the high-Arctic study site. c) Photo of a 3D
- 701 printed snow bunting model in the field at the low-Arctic study site.



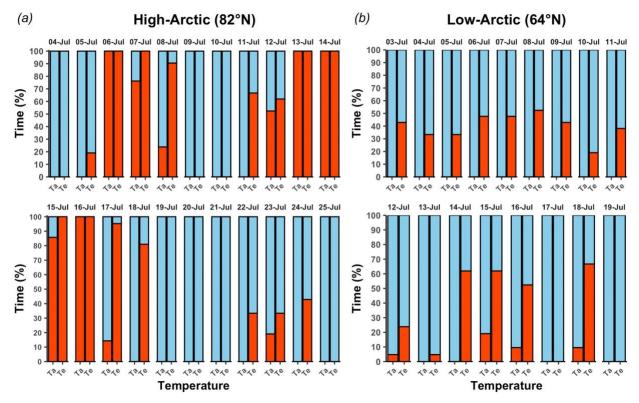
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704 Figure 2. (a) Snow bunting thermoregulatory polygon bounded by basal metabolic rate (BMR; 705 0.564 W), minimum wet thermal conductance (0.023 W/°C), maximum dry thermal conductance (0.073 W/°C), and maximal sustained metabolic rate estimated at 6x BMR in breeding buntings. 706 707 (b) Sustainable performance (expressed as a multiple of BMR) possible for buntings under 708 thermal balance. At operative temperatures below 11.7°C, buntings can maintain thermal balance 709 and sustain optimal performance (i.e., performance  $\geq 4x$  BMR; blue zone). As operative 710 temperatures increase, buntings must reduce activity, and concomitantly metabolic rate, to 711 maintain thermal balance, resulting in a suboptimal performance (i.e., performance < 4x BMR; red zone). Optimal performance is defined as the sustained level of work required by adults to 712 713 sufficiently rear nestlings. The black dashed lines are the extrapolation of the minimum and 714 maximum thermal conductance slopes to the average body temperature recorded during 715 laboratory measurements. The -0.6°C and 24.6°C temperatures on the right side of the polygon 716 in (b) represent the average hourly operative temperature range measured in the field during 717 snow buntings' peak provisioning period.





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733Figure 4. The daily percentage of time during their active period (01:00 hrs to 22:00 hrs) when734snow buntings at a (a) high-Arctic (Alert, Nunavut Canada) and (b) low-Arctic (East Bay Island,735Nunavut Canada) breeding site could either sustain an optimal performance level (blue region) or736forced to work at suboptimal performance levels (red zones) based on either operative (Te) or air737(Ta) temperature recordings. Optimal and suboptimal performance is defined as the periods when738buntings could sustain levels of work  $\geq$  4-times basal metabolic rate or < 4-times basal metabolic</td>739rate, respectively, while maintaining thermal balance under a given heat load.