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- 1 Phenology and predictors of spring emergence for the Timber Rattlesnake (Crotalus
- 2 horridus)
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- 15 Abstract
- Many temperate reptiles survive winter by overwintering in subterranean refugia until external
- conditions become suitable for above-ground activity. Determining when to emerge from refugia
- 18 relies on a reptile's ability to interpret when above-ground environmental conditions are
- 19 survivable. If temperate reptiles rely on specific environmental cues such as temperature to
- 20 initiate egress, we should expect emergence phenologies to be predictable using available local
- 21 climatic data. However, specific predictors of emergence for many temperate reptiles, including
- 22 the Timber Rattlesnake (Crotalus horridus), remain unclear, limiting our understanding of

vertebrates?

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- 23 overwintering phenologies and restricting effective conservation and management. Our
- objectives were to identify potential environmental cues of spring egress for C. horridus to
- 25 determine the species emergence phenology and to examine the applicability 8f identified cues
- 26 in predicting the emergence phenology across the species' range. We used wildlife cameras and
- 27 weather station-derived environmental data to observe and predict the daily surface presence of
- 28 C. horridus throughout the late winter and early spring at communal refugia in Jersey and Jo
- 29 Daviess Counties, Illinois. The most parsimonious model for predicting surface presence
- included the additive effects of maximum daily temperature, accumulated degree days, and
- 31 latitude. With a notable exception in the southeastern U.S., the model accurately predicted the
- 32 average egress day for other populations range wide, emphasizing the role of temperature in
- 33 influencing the substantial phenological plasticity observed across the species' range. The
- 34 apparent broad applicability of the model to other populations suggests it can be a valuable tool
- 35 in predicting spring egress phenology. Our results provide a foundation for further ecological
- 36 enquiries and effective management and conservation strategies.

several other

Introduction

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- 38 Seasonally colder periods in temperate regions expose reptiles to temperatures exceeding critical
- 39 thermal minima (Ganz and Pough, 1982). A common strategy to survive such critically low
- 40 temperatures is to retreat into subterranean refugia until external conditions become suitable for
- above-ground activity. Though buffered from external conditions, occupants of refugia are often
- still subject to thermal regimens lower than preferred (Brown, 1982; Claussen et al., 1991; Firth,
- 43 1998), resulting in a cold-induced dormancy characterized by restricted physiological,
- behavioral, and cellular-level functions (Angilletta and Angilletta, 2009). Cold-induced
- dormancy poses several challenges to reptiles. Despite highly depressed metabolic rates, cold



59	If spring emergence in temperate reptiles is prompted by specific environmental cues such as
70	temperature, we should expect emergence phenologies to be predictable using available local
71	climatic data. Wide-ranging species subject to latitudinal and altitudinal clines might exhibit
72	phenological plasticity, similar to what is observed for insects (Cayton et al., 2015; Herms, 2004;
73	Uelmen et al., 2016), plants (Aslam et al., 2017) and mammals (Boutin and Lane, 2014).
74	However, local adaption or study methodologies have often confounded the identification of
75	such spatial patterns (Andrews and Waldron, 2017; Blouin-Demers et al., 2000). While
76	environmental variables, primarily temperature, likely dictate the timing of emergence, specific ??
77	predictors of egress for many temperate reptiles remain unclear. Furthermore, to our knowledge,
78	no study has examined the applicability of their results across populations, particularly for wide-
79	ranging species exhibiting a high degree of phenological plasticity (Blouin-Demers et al., 2000;
80	DeGregorio et al., 2017; Gregory, 1982; Martin, 1992). The lack of phenological schedules for
81	such species limits our understanding of overwintering ecology and restricts effective
82	conservation and management (e.g., defining date cutoffs for management), particularly for
83	species of conservation concern.
84	The Timber Rattlesnake (Crotalus horridus) is a wide-ranging terrestrial pitviper dependent on
85	subterranean refugia for overwinter survival exhibiting population declines. Dependency on
86	refugia throughout higher latitude within its distribution dictates the species' ecology –
87	exemplified by communal overwintering of up to 200 individuals (Brown, 1993) and seasonal
88	movements between refugia and summer habitat in the fall and spring (Brown, 1992; MacGowan
89 90	et al., 2017; Sealy 2002). Having the largest geographic range of any rattlesnake, C. horridus has the largest geographic range of any rattlesnake, C. horridus exhibits considerable phenological plasticity in its overwintering ecology (Andrews and
91	Waldron, 2017; Brown, 1992; Martin, 2002; Reinert, 2002). Southern populations in warmer-

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46	dormancy creates an energy deficit, and reptiles must survive solely on stored energy reserves.
47	Cold-induced dormancy also restricts a reptile's ability to conduct vital active-season processes
48	such as foraging, reproduction, or basic physiological maintenance (Blouin-Demers et al., 2000;
49	Gregory, 1988; Macartney and Martin, 1993; Smits and Yorke, 1980; Viitanen, 1968). It is,
50	therefore, advantageous for reptiles to limit the duration of cold-induced dormancy and balance
51	the inherent risks of emerging too early and being subject to lethal thermal regimes with the
52	advantage of maximizing active duration out of the refugia.
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53	Determining when to emerge from refugia relies on a reptile's ability to interpret when above-
54	ground conditions are survivable. Such a decision is particularly complicated in temperate
55	climates subject to stochastic environmental conditions. While studies have proposed several
56	cues for egress, including physiological thresholds (Angilletta and Angilletta, 2009), endogenous
57	rhythms (Lutterschmidt, 2006; Weatherhead, 1989), rainfall/humidity (Viitanen, 1967), and
58	photoperiod (Rismiller and Heldmaier, 1982), the most prevalent for temperate reptiles is
59	temperature. Reptiles generally emerge as air temperatures rise in the spring, a phenomenon
60	correlated with several covariates, including maximum, minimum, and mean daily temperatures
61	(Bishop and Echternacht, 2004; Brown, 1992; Graves and Duvall, 1990), accumulated degree
62	days (ADD; Hoffman, 2021; Turner and Maclean, 2022;) and moving "lagged" average
63	temperatures (DeGregorio et al., 2016). Many species emerge only when specific threshold
64	temperatures are surpassed (Blouin-Demers et al., 2000; Burger, 2019; DeGregorio et al., 2017;
65	Sexton and Marion, 1981), although significant inter-individual variation has often confounded
66	the identification of a reliable thermal trigger. Such thresholds likely reflect the thermal
67	dependency of many physiological, behavioral, and cellular-level functions, which dictate the
68	lower thermal limits of surface activity (Angilletta and Angilletta, 2009).

climates emerge up to 3 months earlier (March/April in South Carolina; Andrews and Waldron, 92 2017) than northern populations (May/June in New York; Brown, 1992). 93 Despite studies identifying various temperature-related drivers of egress (Andrews and Waldron, 94 2017; Brown, 1992; Martin, 1992), the range-wide applicability of specific cues remains 95 96 unknown. Consequently, the timing of spring egress for most populations, such as in Midwest states, including Illinois, remains undefined. Such inquiries are particularly relevant for C. 97 horridus throughout northern regions where philopatric individuals congregate at refugia during 98 spring egress and are consequently susceptible to threats including human persecution (Galligan 99 and Dunson, 1979) and management activities such as prescribed burns (Beaupre and Douglas, 100 2012). As a slow-maturing species with infrequent reproductive events (Aldridge and Brown, 101 1995; Bielema, 2022; Brown, 1991), C. horridus lacks the demographic plasticity to recover 102 from population declines rapidly (Brown, 1993), leaving mortality events detrimental to 103 population viability. Therefore, knowledge of spring egress phenology is invaluable for 104 management and conservation purposes.

1. quantity eyess in Illinois 105 Thus, our objectives were to: \nearrow identify potential environmental cues of spring egress for C. 106 107 horridus to determine emergence phenology; and 2) examine the applicability of identified cues in predicting the emergence phenology across the species' range. We used wildlife cameras and 108 weather station-derived environmental data to construct a predictive model for the daily surface accurate? 109 presence of C. horridus during spring egress at refugia across Illinois. Using our model, we 110 predicted the surface presence and egress phenology of other populations reported in other 111 studies and throughout Illinois. Our results detail the overwinter ecology of C. horridus, which 112 provides a foundation for further ecological enquires and effective management and conservation 113

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strategies.

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Materials & Methods

	116	Study Site and Data Collection. — We conducted research at two over-wintering sites located
	117	~350 km (>3° latitude) apart in western Jo Daviess County and Principia College in Jersey
	118	County (Fig. 1). We performed all research under an approved Illinois Department of Natural
	119	Resources Endangered and Threatened Species Permit (#05-11S) and approved University of
	120	Illinois Animal Care and Use Protocols (#22167, #22168). Knowledge of <i>C. horridus</i> at Jo
	121	Daviess County dates to the 1930s, with locations of refugia discovered via visual encounter
	122	surveys in 1991 (Bielema, 2022). Known occupancy at Principia College also dates to the 1930s
	123	with the acquisition of the land for the College, with specific entrances identified and
	124	reconfirmed via visual encounter surveys and VHF-radiotelemetry of individuals beginning in
	125	2010 (S. Eckert, Pers. Coms). Both sites consist of upland mesic forest bounded to the south
	126	(Jersey County) and southwest (Jo Daviess County) by Mississippi River limestone-dolomite
	127	bluffs, covered with a vegetational matrix of remnant hill prairies and oak-hickory dominated
	128	woodlands. Crevices, talus, and holes along the bluff fronts at both sites provide overwintering
	129	refugia for C. horridus.
	130	We monitored <i>C. horridus</i> activity at all identified refugia entrances using Bushnell HD Trophy
	131	Cameras (Model #119736) fitted with external 12V batteries for extended life. Our cameras were
÷	132	in remote, topographically rugged locations away from local communities, and thus no people,
	133	apart from the researchers involved in data collection (the authors), visited the sites or were
	134	photographed. Each camera's position depended on substrate and habitat but was generally set
	135	~1–2 m from the refugia and at the same elevation as the main entrance to afford a satisfactory
	136	field of view. Preliminary investigation revealed cameras occasionally failed to photograph the
	137	slow-moving <i>C. horridus</i> because the passive infrared sensors (PIR) did not detect the snakes.



138	Therefore, we supplemented PIR-triggered photos with the camera's time-lapse feature to take
139	date/time-stamped photographs at 5 min intervals for the deployment period 5 min photo
140	interval provided high-resolution monitoring of refugia while also maximizing battery life and
141	camera uptime. One exception to our camera schedule occurred at Jo Daviess County during
142	spring 2018, where cameras photographed the entrances at 1 hr intervals alongside PIR-triggered
143	photos.
144	Data Analysis. — We visually inspected all photographs for C. horridus and recorded the dates
145	and times of surface presence. We then converted the surface presence of each refugia-year
146	combination into a binomial response variable representing daily surface presence (1 = snake
147	present, 0 = no snake present), omitting days where cameras malfunctioned, and
148	deployment/retrieval days. We determined differences between the days of presence for each
149	refugia-year combination using bootstrapped means and 95% confidence intervals (CIs).
150	Specifically, we resampled the ordinal days of presence for each refugir-year combination with
151	replacement 10,000 times, calculated the mean for each resample, and then determined the mean
152	and 95% CIs based on the resulting bootstrapped resampling distributions (2.5% quantile =
153	lower CI limit; 97.5% quantile = upper CI limit). Non-overlapping CIs indicated informative
154	differences in the effects between refugia-year combinations.
155	We used Generalized Logistic Mixed Effects Models (GLMMs) in the R package 'lme4' (Bates
156	et al., 2014) to examine the effects of environmental variables (Table 1) on the probability of
157	surface presence; the probability of one or more <i>C. horridus</i> being surface active on a given day.
158	For our study, we limited our analysis to variables derived from weather stations <= 30 km from
159	each site, allowing for relative comparisons across different spatial and temporal extents. Due to
160	the prevalence and apparent importance of temperature as a cue for egress within the literature,



we focused primarily on temperature-derived variables found to be important drivers of surface 161 presence for C. horridus and other temperate reptiles: maximum, minimum, and mean daily 162 temperatures, five-day rolling minimum and maximum daily temperature, day of year, latitude, 163 and accumulated degree days of base 5 °C (ADD). ADD is a phenological measure of seasonally 164 increasing cumulative mean daily temperature above a selected threshold temperature, frequently 165 used to predict phenological events and organismal developmental stages for a variety of taxa 166 (Boutin and Lane, 2014; Cayton et al., 2015; Herms, 2004; Uelmen et al., 2016; Hoffman, 2021). 167 We calculated degree days for each day using the formula: $((T_{\text{max}} + T_{\text{min}})/2) - T_{\text{base}}$, where 168 Transcription and Transcriptio 169 threshold (base) temperature. The selection of the base temperature of 5 °C represented the 170 lowest temperature we observed snake surface activity throughout our study. We then summed 171 ("accumulated") the degree day values for each sequential day, starting from 1 January, to 172 calculate ADD over the study period. Before modeling, we Z-transformed (centered and scaled) 173 all variables and tested them for multicollinearity using Variance Inflation Factor analysis, 174 removing highly correlated (VIF >= 5) covariates from the same model. 175 We created a suite of candidate models (Table 2), including a null model (intercept and random 176 effect only) and a fully additive global model based on a priori hypotheses of drivers of surface 177 presence. The dependent variable for each model was the daily presence, the random effect 178 structure was refugia nested within the year, and the fixed effects were a combination of the 179 environmental variables. We included latitude as a fixed effect in all models, serving as a 180 numerical proxy for the site, to examine potential differences between the two populations. Our 181 candidate set also included additive and two-way interactive models (sample size limitations 182 precluded examining models with higher-level interactions) of the same fixed effect 183



184	configurations to account for different hypotheses. For example, a significant interaction
185	between latitude and a temperature-related variable might imply the effect of temperature on
186	surface presence depended on latitude (Jersey County snakes might be surface active at different
187	temperatures than Jo Daviess County). In contrast, an additive model might imply both
188	populations responded equally to temperature, although the probability of surface presence
189	between the two sites might differ. Finally, as demonstrated by other studies, we specified all
190	continuous variables as quadratic terms to account for potential curvilinear relationships
191	(Hoffman, 2021).
192	Examination of candidate models using the R package 'aer' (Kleiber and Zeileis, 2020) revealed
193	no overdispersion, and therefore we ranked all models using Akaike's Information Criterion
194	adjusted for small sample sizes (AICc) in the R package 'AICcmodavg' (Mazzerole and
195	Mazerolle, 2017), and then examined the marginal and conditional effects of the most
196	parsimonious model(s) using the R package 'effects' (Fox and Hong, 2009). Parameters with 95%
197	CIs, not broadly overlapping zero indicated informative predictors of daily presence. We back-
198	transformed the top model for interpretation and graphed the predicted values and 95% CIs using
199	the R package 'ggplot2' (Wickham, 2011). We examined model fit via marginal and conditional
200	coefficients of determination using the R package 'MuMin' (Barton and Barton, 2015).
201	We examined the applicability of the top-ranked AIC _c model in correctly determining the spring
202	egress of other C. horridus populations by comparing our predictions with the average egress
203	dates reported by other studies. Examination of other studies also allowed us to examine the
204	extent of phenological plasticity in spring egress across the species' geographic range. We
205	limited our comparisons to studies providing a detailed assessment of C. horridus egress
206	(Andrew and Waldron, 2017; Brown, 1992; Martin, 2002) instead of briefly mentioning general

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dates with little empirical evidence. For each study, we used our top-ranked AIC_c model to calculate predicted probabilities of surface presence for each ordinal day from day 1 to 243, comfortably spanning the entire egress period at each site, using environmental data gathered from National Oceanic and Atmospheric Association (NOAA) weather stations nearest to the study sites. Because missing data was present in the NOAA datasets, we calculated 15 year "normals" (averages) for estimation using the top model parameters. Doing so afforded complete datasets and determined the typical climatic conditions and probability of surface presence on a given day for each site. Because we were interested in population-level predictions, we held the fixed effect "latitude" at its mean and set the random effects of den and year to zero. If our model predictions were accurate, we expected the day of year with the highest (peak) probability of surface presence at each site to correspond to the average egress day reported by each study. To aid interpretation, we performed a simple linear regression between each study's reported average egress day and latitude and graphed the results with our model predictions. Given our model predictions were accurate, we also used the top-ranked AIC_c model to predict surface presence across the latitudinal gradient of Illinois for each year of the study period (2018-2020), allowing examination of the intra- and inter-annual differences in predicted surface presence across a finer latitudinal scale. We derived the same environmental variables as before from weather stations within each degree of latitude in Illinois (37–42°) and used the top model to generate predicted probabilities of surface presence for each latitude-year combination. As before, because we were interested in population-level predictions, we held the fixed effect latitude at its mean and set the random effects of den and year equal to zero. We determined predicted values and 95% confidence intervals using the R package 'lme4'. Using the 'bootMer' function, we refit the model by resampling the dependent variable, daily presence, with

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replacement 10,000 times, and calculated the predicted values and 95% CIs based on the resulting bootstrapped resampling distributions (2.5% quantile = lower CI limit; 97.5% quantile = upper CI limit). We present graphs of the daily predictions and 14 day moving averages (for examination of general phenological patterns) against ordinal date for each latitude-year combination and averaged across all three years.

Results

We deployed cameras at six overwintering refugia (three in Jersey County and three in Jo Daviess County) for one or more years from 2018–2020 (Table 3), accumulating ~473,000 photos throughout the study. In Jersey County, cameras monitored two refugia for three years and a third den for two years after being discovered in 2019. All refugia in Jersey County were <1km apart and situated on tree-covered talus slopes near bluff prairies. In Jo Daviess County, cameras monitored all three refugia for two years in 2018 and 2019, although we removed all data from one refugium in 2019 because vegetation restricted the cameras' view and obscured observations. Jo Daviess County refugia were < 0.5 km apart and were located on open-canopy outcrops. The dates and durations of camera deployment varied between refugia and years (Table 3), but all deployments successfully spanned the snake emergence periods at their respective sites. Generally, most camera records showed several weeks of no snake presence on either end of the camera deployment periods, although some cameras in Jo Daviess County photographed several post-emergent C. horridus remaining near refugia entrances at the tail-end of the emergence periods. The individuals typically coiled in crevices within the camera's field of view and often occupied the same location daily. We suspect snakes were gravid females who frequently remain What conternal Manuscript to be reviewed for your use?

near refugia after spring emergence until parturition. Because we were interested only in activity 252 related to refugia use, we removed these observations from further analysis (Table 3; Fig. 2). 253 Examination of bootstrapped 95% means and CIs revealed the daily presence of C. horridus at 254 all refugia in Jo Daviess County occurred later in the spring than in Jersey County (Fig. 2). The 255 mean county-level presence day for Jo Daviess County occurred on day 136 (16 May) compared 256 to day 103 (13 April) for Jersey County. The 95% CIs also indicated intra-county differences in 257 presence days between some, but not all, refugia-year combinations in Jersey County (Fig. 2). 258 Early "one-off" surface presence occurred at all refugia in Jersey County in most years (Fig. 2), 259 with the earliest activity occurring on day 55 (24 February). Cameras observed no such early 260 surface presence in Jo Daviess County. Despite early activity, refugia in Jersey County usually 261 exhibited fewer days of surface presence each year (range = 13-27) than in Jo Daviess County 262 (range = 20-41) (Table 3). 263 We used 1525 camera-deployment days in our analysis to predict the surface presence of C. 264 horridus during the late fall and spring from the six refugia (Table 3). Three candidate models 265 received 100% of the AIC_s weights and included additive or two-way interactive effects between 266 267 ADD, maximum daily temperature, and latitude (Table 4). The most parsimonious model included the additive effects between model covariates, accounted for 72% of model weights, 268 and was used for all further analyses. The 95% CIs of ADD and maximum daily temperature in 269 the top model did not span zero, signifying they had strong explanatory power and were strongly 270 related to the surface presence (Table 5). Conversely, latitude narrowly spanned zero, indicating 271 the parameter had weaker explanatory power. 272 The additive-only top model, as opposed to the interactive, implied C. horridus responded 273

equally to maximum daily temperature and ADD. However, the probabilities of surface presence

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275	were higher in Jo Daviess County than in Jersey County for both variables (Fig. 3). The marginal
276	effects of ADD (holding maximum daily temperature constant; Fig. 3A) revealed an increase in
277	the probability of surface presence to a peak at 277.24°C, decreasing thereafter, with the high
278	value of ADD reflecting the accumulation of degree days from day of year 1 (1 January). The
279	marginal effects of maximum daily temperature, holding ADD constant (Fig. 3), revealed that
280	the probability of surface presence increased with increasing. Snakes were not observed in Jo
281	Daviess County when the maximum daily temperature fell below 11°C. In Jersey County, snakes
282	remained present on the surface at temperatures of 5 °C; however, such occurrences represented
283	only 2.5% (4/155) of all days occurring below 11 °C. Therefore, the combined effect of both
284	variables implies the probability of surface presence increases with higher ADD and
285	temperatures (Fig. 3C and 3D).
	how accurately
286	We used seven other studies to examine the applicability of the top AIC _c -ranked model in
	predicted
287	accurately determining spring egress in other C. horridus populations (Table 5). The studies
288	reported the average day of spring egress spanning latitudes from ~32.4° to ~43.8° (Table 5; Fig.
289	6). Simple linear regression (Fig. 6) revealed a later date of spring egress as latitude increased (r ²
290	= 0.81). Our model predicted the average egress day at each site within 10.2 (SD = 13.1) days.
291	Closer inspection revealed our model failed to accurately predict two sites in Hampton (-23 day
292	difference) and Beaufort Counties (-42 day difference), South Carolina (Andrews & Waldron,
293	2017), which inflated prediction error (Fig. 6; Table 6). If we removed the two sites, our model
294	predicted the day of peak egress within 4.6 days (SD = 4.2).
295	We also predicted the probability of surface presence across each latitude-year combination of
296	Illinois (Fig. 6). Visual inspection of each latitude-year subplot suggests the probability of
297	surface presence is highly stochastic throughout late winter and spring, with intra- and inter-year

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phenological differences within each degree of latitude, particularly in more southerly regions. However, a general unimodal trend is apparent, characterized by a steady increase in the probability of surface presence to a peak as individuals egress from refugia and then declining after that as snakes disperse to summer habitats. The peak probability of surface presence between the southernmost (37°; peak probability = day 95) and northernmost Illinois (43°; peak probability = day 137), averaged over the three study years were approximately 42 days apart. Thus, our model suggests a 1° increase in latitude shifts the predicted peak probability of surface presence approximately seven days later into the spring, although substantial annual differences in peak surface presence occurred across latitudes, likely dependent of local climatic variation (Fig. 6). Additionally, comparison between subplots suggests an increased probability of daily surface presence earlier in the season at progressively lower latitudes with early spikes of probability on warmer days, perhaps indicating the greater potential for early "midwinter" Reword. Repetitive. emergences (longer left tails and probability spikes; Fig. 6).

Discussion

We used wildlife cameras and weather station-derived environmental data to successfully observe and predict the daily surface presence of C. horridus throughout the late winter and early spring at communal refugia in Jersey and Jo Daviess Counties, Illinois. The most parsimonious model for predicting surface presence included the additive effects of maximum daily temperature, accumulated degree days (ADD), and latitude. With a notable exception in the southeastern U.S., the model accurately predicted the average egress day for other populations range wide, emphasizing the role of temperature in influencing the substantial phenological plasticity observed across the species' range. The apparent applicability of the model to other populations suggests it can be a valuable tool in predicting the spring egress phenology of

321	unknown populations across much of the species' geographic range, as demonstrated across
322	Illinois. This is a little Misleading. You did predict egress phenology in the Thinois but we don't know whether it was successful in that the model was not tested at other sites in the state.
323	Our results suggest temperature-related variables are strong drivers of spring emergence of C.
324	horridus. ADD allowed our model to capture the general increase in temperature occurring at
325	refugia sites throughout the late winter and early spring. ADD have a long history in
326	phenological predictions of plants (Boutin and Lane, 2014), invertebrates (Cayton et al., 2015;
327	Herms, 2004; Uelmen et al., 2016), and to a lesser extent, reptiles (Hoffman, 2021; Turner and
328	Maclean, 2022). Unlike other time-related variables such as ordinal date and photoperiod
329	(Martin, 1992), ADD allow for flexible predictions of surface presence by accounting for
330	temperature variation across spatial (latitude and altitude) and temporal (years) extents. For
331	example, degree days will accumulate faster in years and regions exhibiting earlier spring
332	warming. Incorporating such climatic variation in phenological studies is particularly important ecologically meanings.
333	for species occupying large geographic ranges subject to varying thermal regimes and exhibiting
334	substantial phenological plasticity.
335	Including maximum daily temperature with ADD allowed our model to capture the highly
336	stochastic thermal regimes characteristic of temperate climates during gradual spring warming.
337	Other studies have implied daily air temperatures are highly influential and suggest surface Citations?
338	activity occurs only once thermal thresholds are surpassed. Spring egress was associated with a
339	maximum air temperature of ~15 °C in New York, South Carolina, and Virginia (Andrews and
340	Waldron, 2017; Brown, 1992; Martin, 1992). Our results concur with the findings, suggesting 15
341	°C corresponds to a 50% probability of surface presence, above which surface presence was
342	more likely than not.

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It is apparent C. horridus only remains surface active until ~11 °C, with some exceptions, perhaps indicating the lower thermal limits of the species and the onset of physiological changes which inhibit surface activity. Agugliaro (2011) showed temperature-dependent metabolic rate depression in C. horridus occurred at 5 °C and 9 °C, with a steep temperature sensitivity in metabolic rate between 9 °C and 13 °C. Similar metabolic sensitivity was found in Red-sided Garter Snakes (Thamnophis sirtalis parietalis) at ~10 °C (Aleksiuk, 1971a, 1976b). Such metabolic responses likely promote energy conservation during cold-induced dormancy and serve as a mechanism to rapidly return to activity with increasing temperatures (Agugliaro, 2011). Snakes tend not to exhibit activity close to their critical thermal limits because of the risks associated with lower performance (Gregory, 1982; Angilletta and Angilletta, 2009). Thus, the warmer temperature of 15 °C may represent the species' voluntary thermal minima, below which most snakes remain within refugia. However, laboratory-based thermal selection studies on C. horridus emphasizing responses to thermal extremes are required to elucidate such thresholds.

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Including latitude afforded us to examine the effects of temperature-related variables between sites and across latitudes. Acknowledging the model included only additive effects, not interactive, between latitude, maximum daily temperature, and degree days is critical for correct interpretation. At both sites, C. horridus responded equally to temperature, but the probability of surface presence was higher in Jo Daviess County than in Jersey County. Such a difference likely reflects the distinct habitat configurations at each site, which influenced the post-emergent behavior of C. horridus and the cameras' subsequent ability to detect surface activity. Specifically, all refugia in Jo Daviess County were on sun-exposed outcrops with abundant crevices and rock shelves, providing a thermally superior basking habitat and protective cover.

	365	Post-emergent C. horridus would frequently use such basking habitat, remaining within the
	366	camera's view and thus increasing the probability of surface presence.
	367	Conversely, refugia in Jersey County were on closed-canopy, loess-covered talus slopes with a
	368	notable lack of undergrowth or rocks near the entrances. Post-emergent <i>C. horridus</i> did not
	369	linger at the entrances but dispersed from the camera's view to nearby "transient" open habitats
	370	such as the bluff front or adjacent hill prairies. Such behavior resulted in fewer daily
	371	observations and a lower probability of surface presence. Thus, our model's observed effect of
	372	latitude implies the probability of observing snake presence via cameras can differ depending on
	373	proximate site characteristics rather than intra-population differences in response to temperature.
	374	With a notable exception in the southeastern U.S., our top model accurately predicted the
	375	average egress day for other C. horridus populations as reported by studies across the species'
	376	geographic range. We should expect some discrepancy between reported versus predicted values
	377	due to potential sources of error, mending the precision of weather station data used for model
	378	predictions and the study's sample size and sampling methods providing egress estimates.
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	380	Despite these potential sources of error, our model provides accurate predictions for most of the sites examined and indicates temperature is largely responsible for the phenological plasticity of spring egress exhibited by <i>C. horridus</i> . Consequently, there is a strong correlation between
	381	spring egress exhibited by <i>C. horridus</i> . Consequently, there is a strong correlation between spring egress and latitude, with spring warming occurring progressively later in northern sites
* 1	382	spring egress and latitude, with spring warming occurring progressively later in northern sites
	383	(Fig. 5), resulting in delayed <i>C. horridus</i> egress. We also see delayed egress at higher, colder
	384	altitudes (Martin, 2002). The reported average mean egress day 135 at a high-elevation site at
	385	1,075 m on the Allegheny Plateau, WV resembles the northern extreme of the species range,

approximately 5° latitude north.



387	Prolonged cold temperatures and subsequent delays in spring emergence at higher latitudes and
388	altitudes (Martin, 1992; Brown, 1992) result in increasingly shorter active seasons with later
389	egress and earlier ingress, directly impacting life histories. Shorter active seasons reduce the time
390	dedicated to foraging, thus limiting yearly energy acquisition and subsequent adjustment of
391	energy budgets between growth, maintenance, and reproduction (Brown, 2016). Consequently,
392	alongside potentially milder temperatures during the active season, C. horridus in colder climates
393	tend to exhibit slower growth rates, delayed sexual maturity, longer intervals between
394	reproductive events, smaller offspring sizes, and lower reproductive success (Aldridge and
395	Brown, 1995; Brown, 2016; Martin, 2002). Ultimately, prolonged cold temperatures are likely
396	responsible for a reduction in overall fitness, as seen in the apparent vulnerability of northern
397	populations, including extirpations from Canada and the northwestern U.S. (Environment
398	Canada, 2010)
399	Our model failed to predict the average egress day at the southernmost two sites in Hampton and
400	Beaufort County (SC) Andrews and Waldron, 2017). We are uncertain about the reasons for the
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	difference but suspect the region's distinct climate might permit an alternate dormancy strategy
402	difference but suspect the region's distinct climate might permit an alternate dormancy strategy that Andrews an wildow (2017)
402 403	difference but suspect the region's distinct climate might permit an alternate dormancy strategy that our model does not account for. The study falls between 32–33° latitude and represents the only
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403 404 405	difference but suspect the region's distinct climate might permit an alternate dormancy strategy that our model does not account for. The study falls between 32–33° latitude and represents the only populations we examined to fall within the humid subtropical climate of the lower East Coast states. Relative to the other study locations, winter temperatures throughout the Lower East Plain? Coast are milder, with average maximum daily temperatures of 15–17 °C in the region's coldest
403 404 405 406	difference but suspect the region's distinct climate might permit an alternate dormancy strategy that our model does not account for. The study falls between 32–33° latitude and represents the only populations we examined to fall within the humid subtropical climate of the lower East Coast states. Relative to the other study locations, winter temperatures throughout the Lower East Coast Coast are milder, with average maximum daily temperatures of 15–17 °C in the region's coldest months of January and February (Andrews and Waldron, 2017). Because of the warmer climate,

410	threshold of daily surface presence, suggesting surface activity of C. horridus could occur on
411	most days throughout the winter months, but they did not comment on any midwinter surface
412	activity.
413	While relatively rare, we observed occasional midwinter surface activity in Jersey county
414	throughout all months except January. We also witnessed the winter emergences of non-target
415	species, including Agkistrodon contortrix, Coluber constrictor, and Thamnophis sirtalis. Other
416	studies note winter emergences in several other snake species, including <i>A. contortrix</i> (Sanders) and Jacob, 1981), <i>C. adamanteus</i> (Stevenson, 2003), <i>C. horridus</i> (Nordberg and Cobb, 2016), C.
417	and Jacob, 1981), C. adamanteus (Stevenson, 2003), C. horridus (Nordberg and Cobb, 2016), C.
418	viridis (Jacob and Painter, 1980), Heterodon platirhinos (Plummer, 2002), and Sistrurus
419	miliarius barabouri (May et al., 1996). Notably, Nordberg & Cobb (2016) identified 60 winter
420	emergence events from 13 C. horridus in Tennessee via indicative spikes in body temperature.
421	However, despite the number of observations, relatively little is known regarding the frequency
422	and ecological significance of winter surface activity in snakes. Presumably, such activity
423	becomes progressively less frequent in higher latitudes where persistently low temperatures
424	restrict surface activity (Brown, 1982; Viitanen, 1967), partially supported by the lack of early
425	winter emergences in Jo Daviess compared to Jersey County and the lack of observed winter
426	activity at other high-latitude sites (Brown, 1992). Furthermore, our predictions of surface
427	activity reveal a progressively higher probability of early surface presence at lower latitudes in
428	Illinois.
429	While we can only speculate on the significance of such winter surface activity without further
430	study, we suspect such activity is the exception and not the rule. Most early emergences in Jersey
431	County were "one-off" events, typically characterized by a single snake emerging and basking at
432	a refugium entrance on warmer days which permitted surface activity. The snakes perhaps





433	attempted to elevate body temperature to fight disease or infection (Nordberg and Cobb, 2016;
434	Clarke et al., 2011; Kluger, 1979). Nordberg and Cobb (2016) observed over 60 emergence
435	events from 12 C. horridus surgically implanted with radio transmitters only a few days before
436	their ingress into refugia. Thus, it is possible the surgical incision sites did not fully heal before
437	the onset of cold-induced dormancy and necessitated above-ground basking. Additionally,
438	Clarke et al. (2011) and Nordberg and Cobb (2016) noted snakes emerging with skin lesions
439	early in the spring were not uncommon and may indicate Snake Fungal Disease. We also
440	observed C. horridus in Jersey County with severe skin lesions and contusions, and although we
441	do not know their causes, they may motivate snakes to emerge and bask. Although the causes
442	behind early emergences are unknown, the surprising frequency of such events warrants further
443	investigation.
444	Conservation implications. — Spring emergence is a vulnerable period for C. horridus,
445	particularly in northerly latitudes where post-emergent and lethargic individuals congregate at
446	communal refugia and are consequently susceptible to local threats, including management
447	activities (Beaupre and Douglas, 2012). Reducing the risk of such threats is vital for effective
448	conservation; for example, scheduling prescribed burn regimes to occur when snakes are less
449	likely to be surface active to reduce potential fire-induced mortality. Yet, the enigmatic nature of
450	C. horridus, paired with the apparent phenological variation across both latitudinal and
451	altitudinal clines, makes determining site-specific spring phenologies difficult and consequently
452	limits conservation. Our model's ability to generate the probability of surface presence for any
453	given day during spring egress is, therefore, a valuable tool for defining conservation and
454	management schedules. However, predictions to new sites should be treated as hypotheses
455	requiring verification through site-specific phenological studies.

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One strategy to direct such schedules is to define specific probability thresholds, which can be translated into dates useful for management and conservation. For example, management activities could be conducted around refugia until the probability of surface presence exceeds a selected threshold. From a management perspective, the probability of surface presence is synonymous with risk; a higher probability indicates a greater potential for snake surface presence and subsequent exposure to management activities. Consequently, the selection of deemed appropriate thresholds depends on the amount of risk willing to be taken given a specific application. We provide a variety of date thresholds for each latitude of Illinois (Table 7), as determined from general probability trends (i.e., 14 day averages across all years; Fig. 6) to aid in conservation and management scheduling. Ideally, harmful activities would occur only when there is minimal risk of snake presence (e.g., probabilities < 5%; Table 7), corresponding to sustained temperatures below the species' suspected thermal limits of 11 °C. However, such thresholds would likely limit management schedules, particularly in milder southern regions (below 39° latitude) where warmer temperatures increase the probability of surface presence earlier in the season. In such cases, effective cutoffs must balance the risk of snake exposure with time allocated to management activities. We also encourage flexible scheduling whenever possible to account for intra-year and latitudinal climatic differences, although we recognize such scheduling would require the frequent calculation of model predictions based on current temperatures, which are not as readily accessible or practical as a single fixed date threshold. Providing managers access to our model to generate up-to-date predictions, for example, via a web portal, could allow for more flexible scheduling.

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An obvious disadvantage of our study methodology is our inability to determine the abundance of surface-active snakes at refugia using only wildlife cameras. Insight into the number of surface-active snakes would afford a more detailed assessment of the spring emergence phenology of C. horridus and the implementation of more effective conservation strategies by incorporating population-level risk assessments. Knowledge of snake abundance would also help differentiate between early "one-off" emergence events by a single snake, particularly in more southern regions, and general spring egress when most snakes emerge and resume active season pursuits. Both differ in associated risk, which we cannot currently distinguish between. Anecdotal observations of the amount of surface activity seen on the cameras (not reported here) suggest the probability surface presence is positively correlated with surface abundance, although such evidence could be misleading as identifying individuals via photographs was not possible. Future research should focus on determining snake abundance to examine populationlevel risk-further, but we are under no illusion obtaining such information for such an enigmatic snake would undoubtedly be time- and energy-intensive, as shown by Brown (1993) and Martin (1993) who spent upward of a decade obtaining such data.

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Figure 1

Location of Timber Rattlesnake ($Crotalus\ horridus$) overwinter sites in West Jo Daviess County (n = 3 refugia) and Principia College, Jersey County (n = 3 refugia), Illinois.

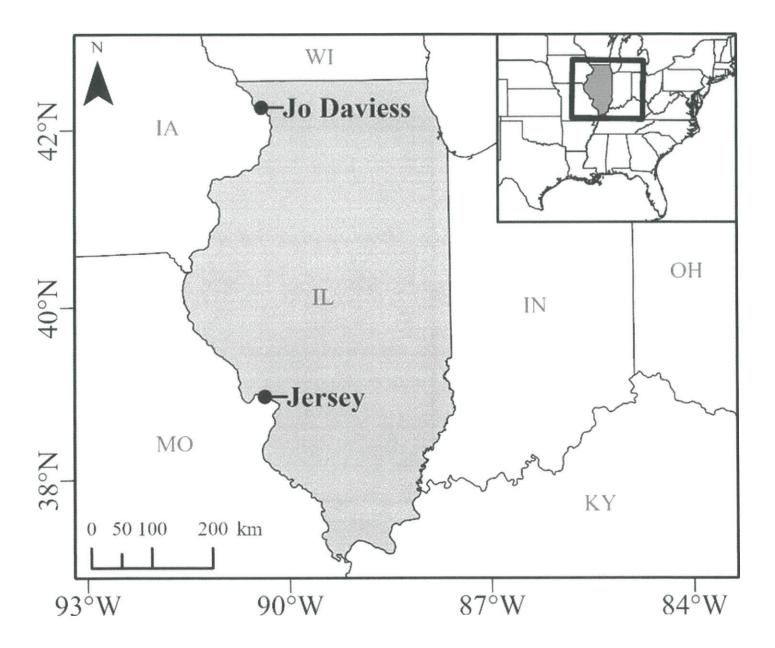


Figure 2

Days of surface presence (grey dots) for Timber Rattlesnakes ($Crotalus\ horridus$) at six communal overwinter refugia in Jo Daviess County (n = 3 dens), and Jersey County (n = 3 dens), Illinois.

We removed days of suspected gravid females (red dots) from further analysis. The mean day of emergence and bootstrapped 95% confidence intervals (solid black dot and error bars) are displayed for each refugia-year combination. *C. horridus* activity observed via Cameras deployed at the entrances of each refugium.

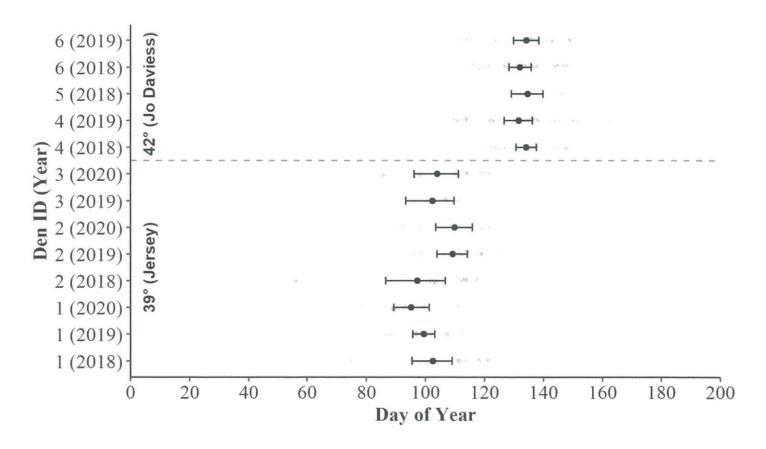




Figure 3

Site-level probabilities and 95% confidence intervals of surface presence by the Timber Rattlesnake ($Crotalus\ horridus$) at communal refugia in Jersey (n = 3) and Jo Daviess (n = 3) Counties, Illinois.

Plots represent: (A) the individual effects of accumulated degree days (ADD) (holding maximum daily temperature constant at its mean of 15.95 °C); (B) maximum daily temperature (holding ADD constant at its mean of 248.88 °C); and (C and D) the additive effects of both variables (in maximum daily temperature increments of 10 °C) for each county. Probabilities determined by the top-ranked AICc candidate model.

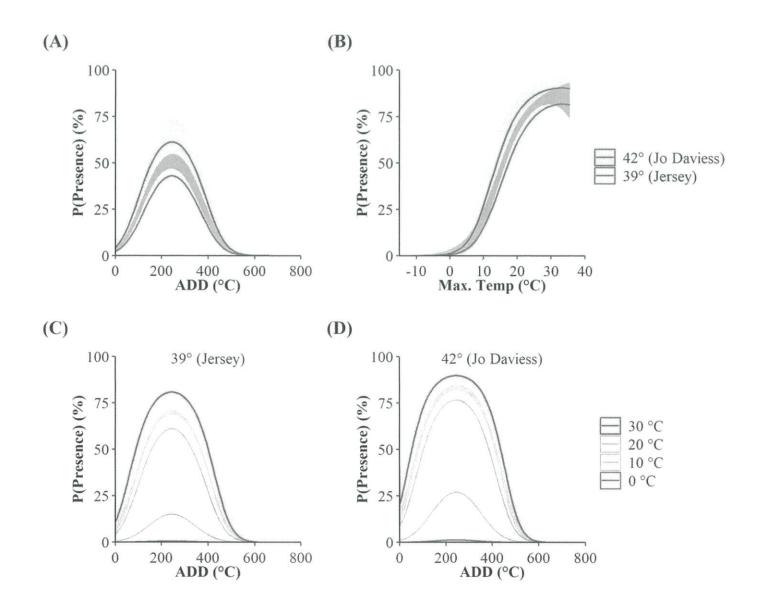




Figure 4

Population-level probabilities (holding latitude constant) of surface presence by the Timber Rattlesnake (*Crotalus horridus*) at communal refugia in Illinois.

Plots represent the individual effects of: (A) accumulated degree days (ADD) (holding maximum daily temperature constant at its mean of 15.95 °C); (B) maximum daily temperature (°C) (holding ADD constant at its mean of 248.88 °C); and (C) the additive effects of both variables (Plots C). Latitude was held constant in all plots (i.e., at the "population"-level). Probabilities were determined by the top ranked AICc candidate model.

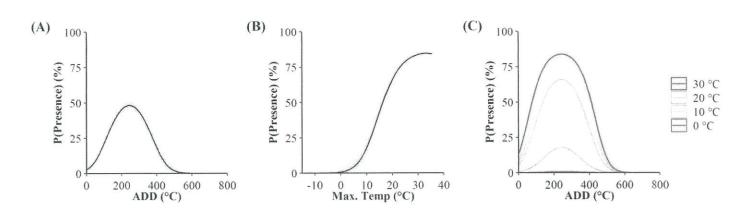


Figure 5

Simple linear regression between latitude and the average day of spring egress, as reported by other studies, overlain with the peak probability of surface presence as predicted by the top AICc model.

Black line and grey ribbon represents estimated values and 95% confidence intervals.

Dashed lines connect each study site's reported egress day (black dots) and predicted probabilities (white dots). Letters correspond to the citation ID of each study detailed in Table 5. Model predictions were calculated using 15 year "normals" (averages) of the model covariates maximum daily temperature and accumulated degree days (ADD; base 5 °C), obtained from the closest weather stations to the site(s) of each respective study.

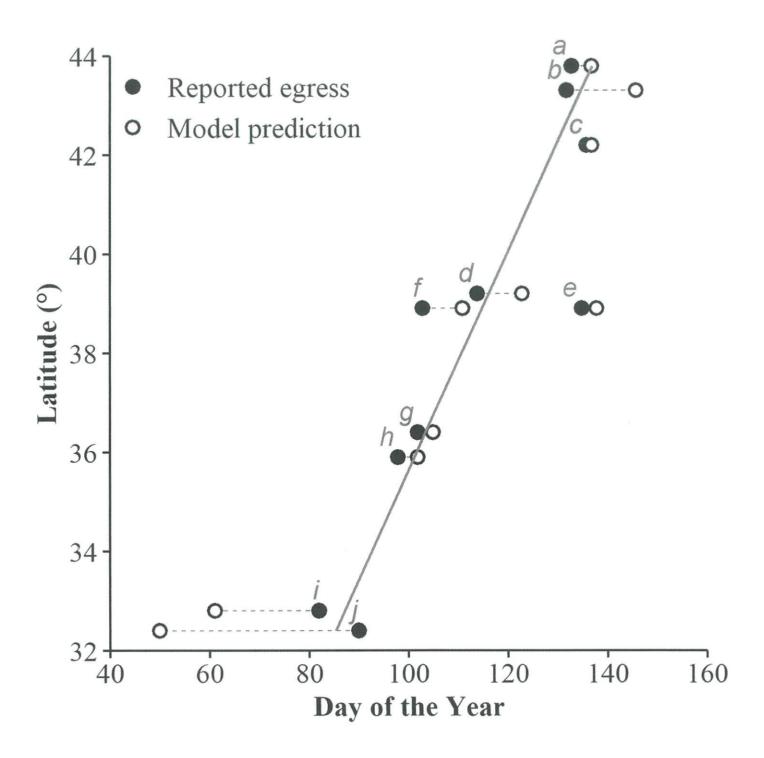


Figure 6

Predicted probabilities of surface presence for the Timber Rattlesnake (Crotalus horridus) during the late winter and spring across the latitudinal gradient of Illinois (in increments of 1°).

Plots are displayed for each year of the study period (2018–2020) and averaged across years. Probabilities are displayed for each day (grey lines) and averaged across a 14 day moving window (black lines) alongside bootstrapped 95% confidence intervals (Grey ribbons). Vertical black lines and parenthesized numbers within each subplot represent the day of peak probability of presence. Predictions derived using the top AICc model examining surface presence as a function of the additive effects of accumulated degree days (ADD; base 5 °C) and maximum daily temperature (°C). Predictions were made on data derived from weather station data located within each degree of latitude.

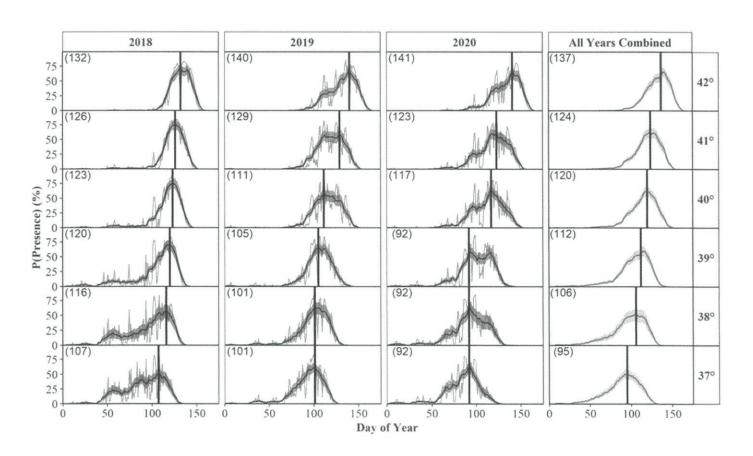




Table 1(on next page)

Environmental variables used to construct candidate models predicting the daily surface presence of *C. horridus*.

Data obtained from weather stations located < 30 km from each field site.



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Predictor	Description (unit)
Min. Temp	Minimum daily temperature (°C)
Max. Temp	Maximum daily temperature (°C)
Mean. Temp	Mean daily temperature (°C)
Min ₅ . Temp	Five-day rolling minimum daily temperature (°C)
Max ₅ . Temp	Five-day rolling maximum daily temperature (°C)
Accumulated Degree Days (ADD)	Cumulative mean daily temperature above 5 °C (°C)
Day of Year	Ordinal date $(1 = January 1^{st})$ (day)
Latitude	Numerical proxy for study site (degrees)



Table 2(on next page)

 AIC_c candidate model set of mixed effects logistic regression models examining the effect of environmental variables on the surface presence of Timber Rattlesnakes (*Crotalus horridus*).

Data collected from refugia in Jo Daviess County (n = 3) and Jersey County (n = 3) during late winter and spring of 2018–2020.

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Table 3(on next page)

Refugia by year deployment summaries of 6 Bushnell HD Trophy Cameras (Model 119736) installed at communal Timber Rattlesnake (*Crotalus horridus*) refugia in Jo Daviess County and Jersey County.

Columns represent camera deployment locations (County, Refugia); Year of camera deployment (Year); dates of camera deployment period (First, Last, Duration); total number of photos taken during the deployment period (Photos); and the number of days *C. horridus* were (Presence Days) or were not (Absence Days) photographed.

	Refugia		Deployment Dates					
County		Year	First	Last	Duration	Photos	Presence Days	Absence Days
		2018	2/24	6/1	98	27,376	27	71
Jersey	1	2019	1/1	6/6	157	69,260	19	138
		2020	1/1	5/31	152	72,002	19	133
	2	2018	2/24	5/31	97	17,914	19	78
Jersey		2019	1/1	6/7	158	46,243	22	136
		2020	1/1	5/31	152	50,230	20	130
Laugary	3	2019	1/1	6/7	158	36,283	13	115
Jersey		2020	1/1	5/29	150	67,045	24	126
I. Davis	4	2018	4/8	6/23	77	7,041	35*	32*
Jo Daviess		2019	3/17	7/13	119	34,165	35*	64*
Jo Daviess	5	2018	4/8	6/13	67	4,567	20	47
Jo Daviess	6	2018	4/8	6/13	67	6,431	38	29
	6	2019	3/17	7/10	116	33,719	41*	47*

^{*}Values represent the number of observed presence days after the removal of days of presumed gravid females (see text).

Change dates

Change dates

Change dates

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Change dates



Table 4(on next page)

 AIC_c results for the top ten mixed effects logistic regression candidate models examining the effect of environmental variables on the surface presence of Timber Rattlesnakes (*Crotalus horridus*).

Results are sorted by ΔAIC_c , where: k = the number of parameters, AIC_c = Akaike score, w_i = Akaike Weights, LL = log-likelihood, R_m^2 = marginal coefficient of determination, R_c^2 = Conditional coefficient of determination

Model	k	AICc	ΔAIC	w_{i}	LL	R ² _m	R^2_c
		788.0		0.7	7/ 2	0.8	0.8
Latitude $+ ADD^2 + Max$. Temp ²	8	3	0.00	2	385.97	3	4
	1	791.2		0.1	-	0.8	0.8
Latitude * Max. $Temp^2 + ADD^2$	0	1	3.18	5	385.53	4	4
	1	791.4		0.1	-	0.8	0.8
Latitude * ADD ² + Max. Temp ²	0	7	3.44	3	385.66	4	4
Latitude $+ ADD^2 + Mean$.		801.1		0.0	-	0.8	0.8
Temp ²	8	9	13.16	0	392.55	3	3
Latitude * Mean. Temp ² +	1	804.1		0.0	-	0.8	0.8
ADD^2	0	2	16.10	0	391.99	3	4
Latitude * ADD ² + Mean.	1	804.7		0.0	1-	0.8	0.8
Temp ²	0	3	16.70	0	392.29	3	3
	1	842.4		0.0	-	0.8	0.8
Global	6	6	54.43	0	412.17	1	2
		850.4		0.0	-	0.8	0.8
Latitude $+ ADD^2 + Min. Temp^2$	8	3	62.40	0	417.17	2	3
	1	852.1		0.0		0.8	0.8
Latitude * Min. $Temp^2 + ADD^2$	0	2	64.09	0	415.99	3	4
	1	853.2		0.0	-	0.8	0.8
Latitude * ADD ² + Min. Temp ²	0	2	65.19	0	416.54	2	3



Table 5(on next page)

Parameter estimates, standard error (SE), and 95% confidence intervals (CI) for top AIC_c model.

The top AICc model examined the additive effects of accumulated degree days (ADD), maximum daily temperature, and latitude on the surface presence of Timber Rattlesnakes (*Crotalus horridus*).



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Parameter	Estimate	SE	Upper CI	Lower CI
Intercept	-4.61	0.36	-5.27	-3.90
ADD	-167.80	17.47	199.26	-133.40
$\mathrm{ADD^2}$	-151.52	13.99	-177.20	-124.71
Latitude	0.36	0.28	0.03	0.63
Max. Temp	80.69	10.20	60.56	99.89
Max. Temp ²	-23.02	8.24	-38.37	-6.53

2



Table 6(on next page)

Summary of studies used to examine the latitudinal variation in spring egress for the Timber Rattlesnake (*Crotalus horridus*), alongside testing our model predictions.

Columns represent the citation ID corresponding to the citation indexes in figure 5 (ID), study citation (Citation); Location of study (Study Location), the year(s) in which research was conducted (Year(s)); The approximate latitude (Lat.) and longitude (Lon.) of the study site (in decimal degrees; the reported average day of egress (Day of Egress); the day of peak probability of surface activity as predicted by the top AIC_c model; and the difference between the reported average day of egress and the day of peak probability of surface presence in days (Diff).



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					Day of	Peak	
ID	Citation	Year(s)	Lat.	Lon.	Egress	Prob.	Diff
a	Brown (1992)	1981-1988	43.8	-73.6	133	137	4
b	Bauder et al. (2011)	2011	43.3	-73.6	132	145	13
c	Current study	2018-2020	42.2	-90.3	136	135	-1
d	Hoffman (2021)	2017-2020	39.2	-82.4	114	122	8
e	Martin (2002)	1989-2001	38.9	-79.3	135	137	2
\mathbf{f}	Current study	2018-2020	38.9	-90.4	103	109	6
g	Sealy (2002)	1990-1997	36.4	-80.3	102	103	1
h	Nordberg & Cobb (2017)	2011-2013	35.9	-86.4	98	100	2
i	Andrews & Waldron (2017)	2002-2004	32.8	-81.1	82	59	-23
j	Andrews & Waldron (2017)	2006-2008	32.4	-80.7	90	48	-42



Table 7(on next page)

The day of the year and corresponding date when the predicted probability of surface presence for *C. horridus* typically exceeded a given threshold in Illinois during our study (2018–2020).

We determined threshold dates for each latitude using 14-day moving averages of predicted probabilities averaged across all years, allowing examination of general phenological trends (Figure 6). We do not report probability threshold values > 60% as 14 day moving averages did not typically exceed this value. Additionally, we do not report probability threshold values of 0% as such occurrences were rare (i.e., there is always some probability of surface presence). Selection of probability thresholds should be determined by the appropriate authorities depending on the specific application (see "Conservation Implications" in the discussion).

	Latitude					
Threshold	37°	38°	39°	40°	41°	42°
5%	54 (2/23)	59 (2/28)	69 (3/10)	87 (3/28)	90 (3/31)	95 (4/5)
10%	56 (2/25)	62 (3/3)	84 (3/25)	93 (4/3)	94 (4/4)	110 (4/20)
15%	65 (3/6)	83 (3/24)	85 (3/26)	94 (4/4)	95 (4/5)	112 (4/22)
20%	80 (3/21)	84 (3/25)	87 (3/28)	95 (4/5)	96 (4/6)	114 (4/24)
25%	82 (3/23)	85 (3/26)	90 (3/31)	96 (4/6)	96 (4/6)	119 (4/29)
30%	85 (3/26)	86 (3/27)	94 (4/4)	97 (4/7)	99 (4/9)	120 (4/30)
35%	86 (3/27)	88 (3/29)	95 (4/5)	98 (4/8)	111 (4/21)	121 (5/1)
40%	90 (3/31)	90 (3/31)	95 (4/5)	99 (4/9)	112 (4/22)	122 (5/2)
45%	91 (4/1)	95 (4/5)	96 (4/6)	100 (4/10)	113 (4/23)	122 (5/2)
50%	92 (4/2)	95 (4/5)	98 (4/8)	111 (4/21)	116 (4/26)	124 (5/4)
55%	95 (4/5)	96 (4/6)	98 (4/8)	112 (4/22)	119 (4/29)	134 (5/14)
60%	96 (4/6)	99 (4/9)	99 (4/9)	114 (4/24)	121 (5/1)	135 (5/15)