1	Predicting the strength of urban-rural clines in a Mendelian polymorphism
2	along a latitudinal gradient
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19	Keywords: Anthropocene; Convergent evolution; Parallel evolution; Urbanization; Selection
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21	Data accessibility: All data and code will be made publicly available on the Dryad digital
22	repository upon publication. All code and data can additionally be found on the GitHub page for
23	JSS (https://github.com/James-S-Santangelo/uac).

25	Acknowledgments: We thank Daniel Anstett, Spencer Barrett, Connor Fitzpatrick, Molly				
26	Hetherington-Rauth, Ruth Rivkin, and John Stinchcombe, whose comments and discussions				
27	greatly improved this project. JSS was funded by an NSERC PGS-D. This work was additionally				
28	funded by NSERC grants to MTJJ and RWN and a Canada Research Chair (CRC) to MTJJ. The				
29	authors declare no conflicts of interest.				
30					
31	Author contributions: JSS conceived of the study with input from KAT and MTJJ. JSS, KAT,				
32	and MTJJ collected samples and JSS, BC, and JS collected data. JSS analyzed the data with				
33	input from MTJJ and RWN. JSS wrote the first manuscript draft with subsequent feedback from				
34	all authors.				
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## 36 Abstract

37 Cities are emerging as models for addressing the fundamental question of whether populations 38 evolve in parallel to similar environments. Here, we examine the environmental factors that drive 39 parallel evolutionary urban-rural clines in a Mendelian trait — the cyanogenic antiherbivore 40 defense of white clover (Trifolium repens). We sampled over 700 urban and rural clover 41 populations across 16 cities along a latitudinal transect in eastern North America. In each 42 population, we quantified the frequency of genotypes that produce hydrogen cyanide (HCN), and 43 in a subset of the cities we estimated the frequency of the alleles at the two genes (CYP79D15 44 and Li) that epistatically interact to produce HCN. We then tested the hypothesis that winter 45 environmental conditions cause the evolution of clines in HCN by comparing the strength of 46 clines among cities located along a gradient of winter temperatures and frost exposure. Overall, 47 half of the cities exhibited urban-rural clines in the frequency of HCN, whereby urban 48 populations evolved lower HCN frequencies. The weakest clines in HCN occurred in cities with 49 the lowest temperatures but greatest snowfall, supporting the hypothesis that snow buffers plants 50 against winter frost and constrains the formation of clines. By contrast, the strongest clines 51 occurred in the warmest cities where snow and frost are rare, suggesting that alternative selective 52 agents are maintaining clines in warmer cities. Additionally, some clines were driven by 53 evolution at only CYP79D15, consistent with stronger and more consistent selection on this locus 54 than on Li. Together, our results demonstrate that both the agents and targets of selection vary 55 across cities and highlight urban environments as large-scale models for disentangling the causes 56 of parallel evolution in nature.

### 57 Impact Summary

58 Understanding whether independent populations evolve in the same way (i.e., in parallel) when 59 subject to similar environments remains an important problem in evolutionary biology. Urban 60 environments are a model for addressing the extent of parallel evolution in nature due to their 61 convergent environments (e.g. heat islands, pollution, fragmentation), such that two distant cities 62 are often more similar to one another than either is to nearby nonurban habitats. In this paper, we 63 used white clover (*Trifolium repens*) as a model to study the drivers of parallel evolution in 64 response to urbanization. We collected >11,000 plants from urban and rural habitats across 16 65 cities in eastern North America to examine how cities influence the evolution of a Mendelian 66 polymorphism for an antiherbivore defense trait – hydrogen cyanide (HCN). This trait had 67 previously been shown to exhibit adaptive evolution to winter temperature gradients at 68 continental scales. Here we tested the hypothesis that winter environmental conditions cause 69 changes in the frequency of HCN between urban and rural habitats. We found that half of all 70 cities had lower frequency of HCN producing genotypes relative to rural habitats, demonstrating 71 that cities drive parallel losses of HCN in eastern North America. We then used environmental 72 data to understand why cities vary in the extent to which they drive reduction in HCN 73 frequencies. The warmest cities showed the greatest reductions in HCN frequencies in urban 74 habitats, while colder, snowier cities showed little change in HCN between urban and rural 75 habitats. This suggests that snow weakens the strength of natural selection against HCN in cities. 76 However, it additionally suggests alternative ecological or evolutionary mechanisms drive the 77 strong differences in HCN between urban and rural habitats in the warmest cities. Overall, our 78 work highlights urban environments as powerful, large-scale models for disentangling the causes 79 of parallel and non-parallel evolution in nature.

## 80 Introduction

81 The extent to which populations adapt in parallel to similar environmental conditions remains a 82 fundamental problem in evolutionary biology (Losos 2017; Bolnick et al. 2018). High levels of 83 genetic and phenotypic parallelism suggest that adaptive evolution is constrained, increasing our 84 confidence in predicting species' responses to similar conditions (Losos 2011). Despite 85 predictions that similar environments should select for similar alleles and phenotypes, the degree 86 of parallelism observed both within and among species is often imperfect (Bolnick et al. 2018). 87 Genetic constraints, genetic drift, and gene flow, among other processes, can all alter the amount 88 of parallelism among populations (Bolnick et al. 2018; Langerhans 2018). Replication is key to 89 disentangling the many causes and consequences of parallel evolution in nature, from 90 macroevolutionary (Schluter 2000) to microevolutionary scales (Lenski 2017), including 91 leveraging naturally repeated cases of adaption across habitat types (Steiner et al. 2009; Stuart et 92 al. 2017; Langerhans 2018). 93 Urban environments are emerging as models for investigating the causes of (non)parallel 94 (sensu Bolnick et al. 2018) evolution among natural populations (Rivkin et al. 2019). Cities tend 95 to share many biotic and abiotic environmental variables such as increased temperatures,

96 elevated pollution, greater habitat fragmentation, and altered structure and composition of

97 ecological communities (McKinney 2006), which can drive parallel adaptive evolution (Reid et

al. 2016; Winchell et al. 2016; Yakub and Tiffin 2016; Kern and Langerhans 2018).

99 Additionally, the commonly observed decreased size and increased isolation of urban

100 populations can drive parallel losses of genetic diversity within urban populations due to stronger

101 genetic drift and restricted gene flow (Munshi-South et al. 2016; Mueller et al. 2018). Despite the

102 many examples of parallel responses to urbanization, imperfect parallelism—wherein

103 phenotypes vary in the direction of evolutionary change between replicate urban and non-urban 104 populations—is also common (Thompson et al. 2016; Diamond et al. 2018), although the causes 105 of non-parallel responses to urbanization are poorly understood (Rivkin et al. 2019). 106 Recent work has used the globally-distributed plant, white clover (*Trifolium repens*) as a 107 model for examining parallel evolutionary responses to urbanization. Thompson et al. (2016) 108 documented repeated reductions in the frequency of HCN—a chemical plant defense against 109 herbivores — within urban populations across three of the four cities examined in northeastern 110 North America. Observational and experimental data show that HCN is predicted to be costly in 111 the presence of cold winter temperatures because the metabolic components of HCN reduce 112 tolerance to freezing (Daday 1954a, 1965; Kooyers et al. 2018). Consistent with this prediction, 113 correlational data suggest that reduced urban snow cover has led to the observed colder winter 114 ground temperatures in some cities relative to rural areas, which drives selection to reduce HCN 115 frequencies in cities (Thompson et al. 2016). The absence of a cline in one of the four previously 116 studied cities was explained by high urban snow depth in both urban and rural locations, which 117 was hypothesized to insulate plants against the damaging effects of frost (Thompson et al. 2016). 118 If urban-rural variation in snow depth is the only cause of urban-rural clines in cyanogenesis, this 119 leads to an explicitly testable prediction: cities lacking snow should lack urban-rural clines in 120 HCN. However, the two previous studies that have documented urban-rural cyanogenesis clines 121 only sampled northern cities where minimum winter temperatures are below freezing (Thompson 122 et al. 2016; Johnson et al. 2018), preventing a reliable test of the hypothesis that colder winter 123 conditions are the primary agent driving the evolution of urban clines in HCN. Sampling cities 124 that vary in winter temperature and frost exposure is required to understand the environmental 125 conditions under which we expect to find (non)parallel responses of HCN to urbanization.

126 Because neutral processes can sometimes drive parallel phenotypic responses in nature 127 (Losos 2011; Bolnick et al. 2018), it is important to reject neutral explanations before inferring 128 that parallel selection has led to repeated adaptive differentiation. Population genetic simulations 129 suggests that although drift could theoretically cause consistent clines if cities experience more 130 drift (Santangelo et al. 2018a), empirical evidence from neutral microsatellite markers does not 131 support this demography, implying selection is the primary driver of clines. (Johnson et al. 2018; 132 Santangelo et al. 2018a), although three additional lines of evidence would help to distinguish 133 between the roles of selection and drift in generating phenotypic clines in HCN. First, although 134 the two-locus epistatic genetic architecture of HCN can lead to the evolution of clines due to drift 135 (Santangelo et al. 2018a), neutral processes are expected to cause allele frequencies to vary 136 randomly at the two underlying loci (Santangelo et al. 2018a). Thus, repeated clines in the same 137 direction at individual loci underlying HCN can only be explained by selection driving 138 differentiation of urban populations (Santangelo et al. 2018a). Second, an absence of clines in 139 warm cities without snow would be consistent with altered selection in urban environments 140 specifically caused by urban-rural gradients in snow depth and minimum winter ground 141 temperatures. Additionally, clines in cyanogenesis to environmental gradients across North 142 American have arisen via sorting of pre-existing and recurrent gene deletions at underlying loci 143 (CYP79D15 and Li) rather than novel mutations, such that multiple deletion haplotypes segregate 144 at both loci in natural populations (Olsen et al. 2013; Kooyers and Olsen 2014; Olsen and Small 145 2018). The presence of multiple deletion haplotypes in both rural and urban populations would 146 strengthen our inference that cyanogenesis is the target of selection rather than specific genes 147 linked to the HCN loci; selection on linked genes are expected to lead to a single haplotype being 148 overrepresented in urban or rural populations.

149 Here, we combine sampling of over 11,000 white clover plants from 16 cities with 150 publicly available climate data to assess the environmental drivers underlying the evolution of 151 latitudinal and urban-rural clines in cyanogenesis across multiple cities in eastern North 152 America. We begin by assessing the environmental predictors of HCN frequencies along the 153 latitudinal gradient by asking: (1) What regional environmental factors predict mean HCN 154 frequencies within populations? Consistent with previous work (Daday 1954a, 1965, Kooyers 155 and Olsen 2012, 2013), we expected to observe lower HCN frequencies at more northern 156 latitudes due to lower winter temperatures. We then examine urban-rural clines in HCN across 157 16 cities to address the following questions: (2) How common are urban-rural clines among large 158 (> 2 million people) cities in eastern North America? (3) What regional environmental factors 159 predict the strength of clines in cyanogenesis? We predicted that we would observe the weakest 160 clines in cities with high minimum winter temperature (i.e., warm cities) and also in those with 161 high levels of snowfall due to weaker frost-mediated selection against HCN-producing 162 genotypes. (4) Are clines present at both genes underlying cyanogenensis? Repeated clines at 163 both loci underlying HCN would suggest that genetic drift does not cause urban-rural clines in 164 HCN and that selection specifically acts on the production of HCN, as opposed to alternative 165 functions of the individual loci. Finally, we ask: (5) Do urban populations show only a subset of 166 the variation in deletion haplotypes as rural populations? The presence of multiple or fewer 167 deletion haplotypes segregating in urban populations relative to rural populations would suggest 168 that selection favors acyanogenic genotypes directly, rather than linked sites. Our results 169 highlight urban environments as large-scale, replicated systems for addressing the ecological and 170 genetic underpinnings of (non)parallel evolutionary responses in nature.

#### 172 Materials and methods

### 173 Study system

174 *Trifolium repens* (Fabaceae) is a perennial legume that reproduces clonally through the

- 175 production of stolons and sexually through self-incompatible, hermaphroditic flowers arranged
- 176 in dense inflorescences (Burdon 1983). Plants are typically found in grazed or mowed pastures,
- 177 lawns and meadows where they can maintain large dense populations (Burdon 1983). Native to
- 178 Eurasia, T. repens was introduced to temperate regions worldwide as a forage and nitrogen-
- 179 fixing crop (Burdon 1983; Kjærgaard 2003). Because of its long history of human-mediated
- 180 transport, white clover is found in cities all over the world, making it an ideal system for
- 181 studying patterns of parallel evolution in response to urbanization.

182 Many white clover populations are polymorphic for the production of hydrogen cyanide 183 (HCN), with cyanogenic (HCN present) and acyanogenic (HCN absent) cyanotypes co-occurring 184 (Daday 1958). The molecular and genetic basis at the individual loci underlying the production 185 of both metabolic components involved in HCN production was recently characterized (Olsen et 186 al. 2007, 2008, 2013; Olsen and Small 2018). The Ac/ac polymorphism is caused by deletions 187 overlapping the CYP79D15 locus (hereafter Ac), which encodes the cytochrome P450 subunit 188 involved in the synthesis of cyanogenic glycosides (linamarin and lotaustralin) stored in the cell 189 vacuole (Olsen et al. 2008, 2013; Olsen and Small 2018). Plants require at least one functional 190 allele with dominant expression (i.e., Ac-) to produce cyanogenic glycosides. Similarly, the Li/li191 polymorphism results from a deletion at the *Li* locus encoding the hydrolyzing enzyme 192 linamarase, which is stored in the cell wall (Kakes 1985); at least one dominant allele (i.e., Li-) 193 is required to produce linamarase. Thus, plants require a minimum of one dominant allele at each 194 locus to produce HCN (i.e., cyanotype Ac-Li), which is released when cell damage causes

cyanogenic glycosides and linamarase to interact (Hughes 1991). If either locus is homozygous
for the recessive allele, then a plant lacks HCN and is said to be "acyanogenic" (i.e., cyanotypes *Ac–lili, acac Li–, acac lili*).

198

199 Sampling and HCN assays

200

201 In May and June 2016, we sampled 15 plants from each of 15 to 45 populations (mean =  $\sim$ 38) 202 along urban-rural gradients in each of 12 cities in eastern U.S.A (Fig. 1). Power analyses 203 conducted by resampling the data from Thompson *et al.* (2016) showed that this sampling 204 scheme provides sufficient power to detect even the weakest statistically significant clines in 205 HCN (see supplemental text S2: "Power analyses for sampling design", Fig. S1). We sampled 206 only large cities (240,000 < Human population size (city area) < 8,200,000; 151 < city area (km<sup>2</sup>) 207 < 2,300.) because these are likely to have the strongest environmental gradients associated with 208 urbanization. We additionally chose cities along a north-south latitudinal transect such that more 209 southern cities had less snow and warmer winter ground temperatures, which earlier research 210 suggested would weaken selection against HCN in urban environments, leading to weaker or 211 absent clines in southern cities (Thompson et al. 2016). Sampling took place in three trips: trip one (May 16<sup>th</sup> to 23<sup>rd</sup>, 2016) involved collections 212

from Tampa and Jacksonville, FL, Atlanta, GA, and Charlotte, NC; trip 2 (June 5<sup>th</sup> to 11<sup>th</sup>, 2016) from Norfolk, VA, Washington, D.C., and Baltimore, MD, and Philadelphia, PA; and trip 3 (June 15<sup>th</sup> to 20<sup>th</sup>) from Cleveland and Cincinnati, OH, Pittsburgh, PA, and Detroit, MI. In each city, we targeted populations spaced at least 1 km apart. In each population, we recorded the latitude and longitude coordinates and collected ~6 cm-long white clover stolons with three to

218	four intact leaves; stolons were at least 1.5 m apart to minimize sampling the same clonal
219	genotype. Stolons were placed in sandwich bags and kept on ice in a cooler until being brought
220	back to the lab where they were individually placed in 2 ml microcentrifuge tubes and stored at
221	-80 °C until HCN phenotyping. In total, we collected and assayed 6,738 stolons from 459
222	populations across 12 cities. For all downstream analyses, we combined the data from the 12
223	cities described above with the 4 cities (Boston, MA, New York, NY, Toronto and Montréal,
224	CA) originally sampled by Thompson et al. (2016). In total, we analyzed urban-rural clines in
225	HCN using 11,908 plants from 721 populations across 16 cities.
226	We used Feigl-Anger assays to determine whether plants were cyanogenic or
227	acyanogenic (Feigl and Anger 1966; Gleadow et al. 2011), which uses a simple color change
228	reaction to determine the plant's phenotype. Briefly, we added a single mature leaf to wells in
229	96-well plates with 80 $\mu$ L of sterile deionized water. Leaf samples were added to alternating
230	wells so that a single plate could accommodate 48 plant samples. The plates were frozen at
231	-80°C to facilitate cell-lysis and release of HCN, and upon removal we macerated the plant
232	tissue with pipette tips. We then covered the plate with Feigl-Anger test paper and incubated the
233	covered plate at 37°C for 3 hr. Cyanogenic individuals (i.e., Ac-Li-) produce a blue spot on the
234	filter paper above the well, whereas acyanogenic plants (i.e., Ac-lili, acac Li-, or acac lili)
235	produce no color change.
236	To assess whether clines were driven by changes in the frequency of HCN or clines at

either component gene (i.e. *Ac* or *Li*), we determined the frequency of *Ac* and *Li* for a subset of
the cities (Atlanta, Baltimore, Charlotte, Cleveland, Jacksonville, New York, Norfolk, and
Washington), which was combined with previously-collected allele frequency information for *Ac*and *Li* for the city of Toronto (Thompson et al. 2016). For plants that tested negative for HCN,

241	we added either (1) 30 $\mu$ L of 10 mM exogenous cyanogenic glycosides (linamarin, Sigma-
242	Aldrich 68264) plus 50 uL of ddH_20 or (2) 80 $\mu L$ of 0.2 EU / mL linamarase (LGC Standards
243	CDX-00012238-100). A positive reaction in (1) indicates a plant producing linamarase (i.e. <i>acac</i>
244	<i>Li</i> –); a positive reaction in (2) indicates a plant producing glycosides (i.e. <i>Ac</i> – <i>lili</i> ); a negative
245	reaction in both indicates plants that do not produce glycosides nor linamarase (i.e. the double-
246	homozygous recessive genotype, acac lili). These assays have been previously confirmed
247	through PCR to reliably determine the cyanotype of individual <i>T. repens</i> genotypes (Olsen et al.
248	2007, 2008; Thompson and Johnson 2016). Due to the complete dominance of functional alleles
249	at both loci, we are unable to calculate the frequency of $Ac$ or $Li$ solely from the phenotyping
250	assays described above (e.g. AcAc and Acac are indistinguishable). We therefore used the marker
251	frequency (e.g. Ac-) to calculate the frequency of Ac and Li assuming Hardy-Weinberg
252	equilibrium $(p^2 + 2pq + q^2 = 1)$ , where $q^2$ represented the observed frequency of homozygous
253	recessive genotypes at Ac (i.e. acac) or Li (i.e. lili). While urban-rural HCN clines may not
254	always meet the assumptions of HWE (Johnson et al. 2018; Santangelo et al. 2018a), deviations
255	from HWE are not predicted to greatly impact inferred allele frequencies when homozygous
256	dominant and heterozygous individuals are phenotypically identical, as is the case for HCN
257	(Lachance 2009; Kooyers and Olsen 2013). Analyzing changes in the frequency of Ac and Li for
258	these nine cities allowed us to assess whether selection is targeting HCN specifically or
259	individual loci underlying HCN production.
260	To examine whether urban and rural populations varied in the frequency of deletion

haplotypes at Ac or Li, we used PCR with previously described forward-reverse primer pairs (Olsen et al. 2014) to identify the relative size of deletions at both loci for individual plants. We extracted total genomic DNA from each of 10 randomly-selected urban and rural plants (n = 20)

264 for each of seven cities (Table 1) using a standard CTAB-chloroform extraction method 265 (Agrawal et al. 2012). We chose these cities because they spanned the range of latitudes included 266 in our study, thus reducing potential impact of geographical variation on haplotype frequencies. 267 We included cities that varied in the presence (5 cities) and absence (2 cities) of clines in HCN. 268 We only extracted DNA from plants that were homozygous recessive at both loci (i.e. acac lili) 269 because these plants have at least one deletion haplotype at each locus. Each plant was amplified 270 with 6 different primer pairs (3 for each locus), designed to assay the approximate size of the 271 genomic deletion at each locus based on the presence/absence of PCR products (Kooyers and 272 Olsen 2014). Note that larger deletions are masked by smaller deletions when resolving 273 haplotypes on a gel, preventing us from estimating the true frequency of each deletion; we 274 therefore rely only the presence/absence of deletions in our analyses (see "Statistical analyses") 275 below). Using this approach, we were able to assign 88% and 78% of samples to previously 276 described Li (n = 4) and Ac (n = 2) deletion haplotypes, respectively. The remaining individuals 277 were either newly discovered haplotypes or individuals with intact Li and Ac genes, the latter 278 indicating either false negative phenotyping or false positive haplotyping assays (or the presence 279 of a silencer modifier locus). We focus our results on the haplotypes that aligned with those 280 previously described by Kooyers and Olsen (2014) to understand whether one 281 deletion haplotype versus multiple haplotypes were segregating in urban and rural populations in 282 cities across our latitudinal transect. 283

284 Environmental data

To examine the regional abiotic factors that predict the strength of clines in HCN, we retrieved environmental data from publicly available databases. First, we retrieved the minimum winter

287	temperature (MWT, Bio6) — an important predictor of HCN frequencies (Daday 1965; Kooyers
288	and Olsen 2013) — and the maximum summer temperature (MST, Bio5) using the highest
289	resolution data (30 arc seconds; 1 km <sup>2</sup> ) available from the BioClim database (version 1.4,
290	Hijmans et al. 2005). We additionally retrieved the average monthly precipitation (Precip) at 1
291	km <sup>2</sup> resolution from the same database. Next, we obtained the annual aridity index (AI), monthly
292	average potential evapotranspiration (mPET) and average annual potential evapotranspiration
293	(aPET) at 1 km <sup>2</sup> resolution from the Consortium for Spatial Information (CGIAR-CSI, Trabucco
294	and Zomer 2009). These three abiotic factors are known predictors of the frequency of HCN and
295	its component genes at the continental scale in North America, Europe and New Zealand
296	(Kooyers and Olsen 2013). BioClim and CGIAR datasets are provided as gridded raster layers,
297	from which we extracted the relevant data for all 721 populations using QGIS v. 3.2.3 (QGIS
298	Development Team 2018).
299	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum
299 300	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum temperature for all cities for the years 1980 to 2015 from the National Oceanic and Atmospheric
299 300 301	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum temperature for all cities for the years 1980 to 2015 from the National Oceanic and Atmospheric Administration's National Centers for Environmental Information
<ul><li>299</li><li>300</li><li>301</li><li>302</li></ul>	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum temperature for all cities for the years 1980 to 2015 from the National Oceanic and Atmospheric Administration's National Centers for Environmental Information (https://www.ncdc.noaa.gov/cdo-web/datatools/selectlocation). Importantly, these are regional
<ol> <li>299</li> <li>300</li> <li>301</li> <li>302</li> <li>303</li> </ol>	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum temperature for all cities for the years 1980 to 2015 from the National Oceanic and Atmospheric Administration's National Centers for Environmental Information (https://www.ncdc.noaa.gov/cdo-web/datatools/selectlocation). Importantly, these are regional environmental data obtained from a single weather station for each city located at the nearest
<ul> <li>299</li> <li>300</li> <li>301</li> <li>302</li> <li>303</li> <li>304</li> </ul>	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum temperature for all cities for the years 1980 to 2015 from the National Oceanic and Atmospheric Administration's National Centers for Environmental Information (https://www.ncdc.noaa.gov/cdo-web/datatools/selectlocation). Importantly, these are regional environmental data obtained from a single weather station for each city located at the nearest international airport; thus, these data represent city-level abiotic conditions and not the
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<ul> <li>299</li> <li>300</li> <li>301</li> <li>302</li> <li>303</li> <li>304</li> <li>305</li> <li>306</li> <li>307</li> <li>308</li> </ul>	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum temperature for all cities for the years 1980 to 2015 from the National Oceanic and Atmospheric Administration's National Centers for Environmental Information (https://www.ncdc.noaa.gov/cdo-web/datatools/selectlocation). Importantly, these are regional environmental data obtained from a single weather station for each city located at the nearest international airport; thus, these data represent city-level abiotic conditions and not the conditions extracted for each population. Some filtering and processing of the environmental data was required prior to downstream analyses. First, we took the mean MWT (°C), MST (°C), AI, and aPET across all populations within a city to estimate the city-level minimum winter temperature, maximum
<ul> <li>299</li> <li>300</li> <li>301</li> <li>302</li> <li>303</li> <li>304</li> <li>305</li> <li>306</li> <li>307</li> <li>308</li> <li>309</li> </ul>	Finally, we obtained daily snow depth, snowfall, maximum temperature, and minimum temperature for all cities for the years 1980 to 2015 from the National Oceanic and Atmospheric Administration's National Centers for Environmental Information (https://www.ncdc.noaa.gov/cdo-web/datatools/selectlocation). Importantly, these are regional environmental data obtained from a single weather station for each city located at the nearest international airport; thus, these data represent city-level abiotic conditions and not the conditions extracted for each population. Some filtering and processing of the environmental data was required prior to downstream analyses. First, we took the mean MWT (°C), MST (°C), AI, and aPET across all populations within a city to estimate the city-level minimum winter temperature, maximum summer temperature, aridity, and annual potential evapotranspiration, respectively. Next, we

310 calculated an alternative measure of aridity, the soil moisture deficit (SMD), as the difference 311 between Precip and mPET. This measure of aridity can more reliably estimate regional aridity in 312 cases where both precipitation and PET are low (Thompson et al. 2014). We calculated SMD for 313 all months from May to September to represent the plant growing season, and again took the 314 mean of these measurements across all populations as our measure of city-level SMD. Finally, 315 we used NOAA's weather station data to estimate regional snow depth (cm), snowfall (cm), and 316 the number of days below zero with no snow cover, a measure of frost exposure that has been 317 previously associated with urban-rural clines in HCN (Thompson et al. 2016). To retrieve these 318 estimates, we first filtered the weather data to remove missing data and only included 319 observations from January and February as these are the coldest winter months in eastern North 320 America. We additionally removed observations from years where data was unavailable for both 321 January and February and eliminated months with fewer than 10 days of weather data. Following 322 filtering, we retained 31,005 weather observations, with a mean of 1937 observations per city 323 (Table S1). From these data, we calculated the mean snow depth, mean snowfall, and the number 324 of days below 0  $^{\circ}$ C with no snow cover (i.e. snow depth of 0 cm) and took the mean of these 325 measurements across all years as our estimates of regional winter conditions. 326

327 Statistical analyses

328

329 For brevity, we only briefly describe the statistical procedures used throughout the paper; a

detailed description of all analyses can be found in the supplementary materials (text S1:

331 "Detailed statistical analyses"). We first tested whether, on average, cities varied in mean HCN

332 frequencies and whether urbanization influenced HCN frequencies. To do this, we fit an

333 ANOVA using type III SS with within-population HCN frequencies as the response variable and 334 city, standardized distance to the urban center and the city × distance interaction as predictors. 335 We used distance to the urban center as a measure of urbanization as this is highly correlated with % impervious surface ( $R^2 = 0.64$ , Johnson et al. 2018) and sufficiently captures variation in 336 337 HCN frequencies across urban-rural gradients (Thompson et al. 2016; Johnson et al. 2018). 338 Because urban-rural transects varied in length, we scaled distance within cities between 0 (most 339 urban) and 1 (most rural). In our model, a significant effect of City suggests that mean HCN 340 frequencies vary across the 16 cities. A significant Distance term means that across all cities, 341 HCN frequencies vary in parallel across the urban-rural transect (i.e. parallel clines in HCN 342 frequencies), while a significant  $\text{City} \times \text{Distance interaction indicates the strength or direction of}$ 343 clines in HCN varies across cities. The significant effects of City and the City × Distance 344 interaction in our model (see "Results") justify an examination of the environmental predictors 345 of mean HCN frequencies and variation in the strength of clines across cities, respectively. 346 To assess the environmental predictors of mean HCN frequencies across cities along our 347 latitudinal gradient, we fit the following linear model: mean HCN frequency ~  $PC1_{HCN}$  + # days 348  $< 0^{\circ}$ C with no snow + annual aridity index + soil moisture deficit, where PC1<sub>HCN</sub> is a composite 349 axis generated via PCA that explained 90.2% of the variation in MST, MWT, summer 350 precipitation, annual PET, and snowfall, all of which were highly correlated and individually 351 significantly predicted variation in HCN frequencies. Lower values of PC1<sub>HCN</sub> represented cities 352 with higher summer temperatures, higher minimum winter temperatures, higher summer 353 precipitation, greater potential evapotranspiration, and lower snowfall. We assessed significance 354 of model predictors using an AIC<sub>c</sub>-based multi-model selection and averaging process.

355 Upon confirming that cities varied significantly in the strength of urban-rural phenotypic 356 clines using ANOVA, we used a similar approach to that described above to examine the 357 environmental predictors of the strength of urban-rural phenotypic clines in HCN. For each city, 358 we first fit a linear regression with the proportion of cyanogenic plants within each population as 359 the response variable and standardized distance to the urban center as the sole predictor. Note 360 that cities that showed significant changes in HCN frequency with distance (Table 1) were also 361 significant following Bonferroni correction of logistic regressions using data from individual 362 plants (i.e., 1 for HCN+, 0 for HCN-). We extracted the slope (i.e.  $\beta$  coefficient) from each 363 city's model as a measure of the strength of the clines and examined the environmental 364 predictors of cline strength by running the following model:  $\beta \sim PC1_{slope}$ , where PC1<sub>slope</sub> is a 365 composite axis generated via PCA that explains 92.8% of the variation in snow depth, snowfall, 366 MWT, and MST, all of which were highly correlated and on their own significantly predicted 367 variation in the strength of clines (see text S1). Cities with low values along PC1<sub>slope</sub> experience 368 little snow, higher minimum winter temperature, and higher maximum summer temperature. 369 We explored whether clines were present at each of the two loci underlying HCN. This 370 was done by fitting linear models in which the allele frequencies for the Ac and Li loci were 371 treated individually as a response variable in separate analyses, which was regressed against 372 standardized distance to the urban core as a predictor. Finally, to examine variation in deletion 373 haplotypes across urban and rural habitats, we used the raw counts of deletion haplotypes at each 374 locus to calculate Simpson's diversity index for deletions in urban and rural habitats for each 375 city, which accounts for both presence/absence and abundance of haplotypes in each population 376 when estimating diversity. We fit Simpson's diversity index as the response variable in a linear 377 model with habitat type (i.e. urban vs. rural) as the sole predictor such that a significant effect of

378 habitat suggested differences in deletion haplotype diversity among urban and rural habitats. All

380

- 381 Results
- 382
- 383 Variation in HCN frequencies in cities along a latitudinal gradient

384

385 The mean frequency of HCN varied across the 16 cities from 19% (New York) to 99% (Tampa) 386 (Effect of City:  $F_{15,689} = 18.48$ , P < 0.001, Table 1, Fig. 1, Fig. 2), with the highest frequencies at 387 the most southern latitudes (Fig. S2). The number of days  $< 0^{\circ}$ C with no snow cover and PC1<sub>HCN</sub> 388 together accounted for 94.1% of the variation in mean HCN frequencies among cities (Table S4). 389 Specifically, HCN frequencies decreased by 1.5% for every additional day below 0 °C with no 390 snow cover ( $\beta = -0.015$ , z = 8.77, P < 0.001, Table S5, Fig. 3a), and by 6.4% for every unit 391 increase along PC1<sub>HCN</sub> ( $\beta = -0.064$ , z = 7.0, P < 0.001, Table S5, Fig. 3b), suggesting HCN 392 frequencies decrease in colder, wetter environments that get more snow. Annual aridity index 393 was not significant predictor of mean HCN frequencies in our model (P = 0.36, Table S5) while 394 soil moisture deficit was not included in any top models following model selection and 395 averaging. 396 397 Environmental predictors of urban-rural clines in HCN frequencies

398

399 On average, urbanization was associated with reduced HCN frequencies across cities, whereby

400 the main effect of standardized distance from the urban center was positively associated with the

analyses were performed in R v. 3.6.0 (R Core Team 2019).

401 frequency of HCN within T. repens populations (main effect of Distance,  $F_{1.689} = 36.42$ , P < 100402 0.001, Fig. 2). In a model with unstandardized distance as a predictor, this translated into an 403 average increase in HCN frequency of 0.3 % per km from the urban center. However, the 404 strength of urban-rural phenotypic clines in HCN varied across cities (Distance  $\times$  City 405 interaction:  $F_{15,689} = 3.26$ , P < 0.001, Table 1, Fig. 2). The strength of urban-rural clines decreased with increasing values along PC1<sub>Slope</sub> ( $R^2 = 28.4\%$ ,  $\beta = -0.036$ ,  $t_{13} = -2.27$ , P = 0.04, 406 407 Table 2, Fig. 4), implying that the strongest clines occurred in the warmest environments and the 408 weakest clines occurred in regions of low temperature and high snowfall/depth. 409 410 *Clines at loci underlying HCN and deletion haplotype frequencies* 411

412 Of the 16 cities surveyed in this study, we assayed the genotype at the two underlying genes in 413 nine cities, six of which showed significant clines in HCN (Table 1). Of the six cities with 414 significant clines, three (Atlanta, Jacksonville, Toronto) showed significant linear clines at both 415 Ac and Li, three (New York, Norfolk, Washington) showed significant linear clines only at Ac, 416 whereas no cities had a significant linear cline only at Li. Significant clines at Ac and Li were 417 always in the same direction as clines in HCN (i.e., lower frequencies of the dominant alleles Ac 418 and Li in urban populations). None of the three cities that lacked a cline in HCN showed a cline 419 at either underlying gene.

420 All deletion haplotypes at *Ac* and *Li* identified previously in this system were found in 421 each city, and their relative frequencies did not vary for either locus between urban and rural 422 populations (fig. S3 and S4). Haplotype diversity (*Ac*: Simpson's  $D_{rural} = 0.41$ ,  $D_{urban} = 0.38$ ,  $t_{11} =$ 423 -0.23, P = 0.82; *Li*: Simpson's  $D_{rural} = 0.60$ ,  $D_{urban} = 0.52$ ,  $t_{11} = -0.74$ , P = 0.47) did not vary 424 across urban and rural habitats for either locus. Together, these results suggest that no specific
425 deletion haplotype is favored in urban habitats at either *Ac* or *Li*.

426

### 427 Discussion

428 We combined field sampling of white clover populations from large eastern North American 429 cities with environmental data to assess the environmental predictors of cyanogenesis on a 430 continental scale and of urban-rural gradients in HCN frequencies. Several key results are most 431 important to answering our research questions. As expected, HCN frequencies decreased 432 northward across the continent as frost exposure increased (question 1). Urban-rural 433 cyanogenesis clines occurred in half of the cities studied (question 2) and the strongest clines 434 occurred in the warmest environments (question 3). Clines in HCN were matched by clines at 435 one or both of the loci underlying HCN, and these clines were always in the same direction 436 (question 4). Finally, the diversity of deletion haplotypes among acyanogenic plants was 437 consistent across urban and rural populations of multiple cities (question 5). Together, these 438 results provide compelling evidence that selection is driving parallel evolution of cyanogenesis 439 clines across multiple large urban centers in North America, although regionally cold and snowy 440 climates dampen parallel responses of HCN to urbanization. Below, we discuss our results in the 441 context of the environmental drivers of HCN evolution at the scale of entire continents and 442 individual cities.

443

#### 444 Environmental predictors of HCN frequencies

The cyanogenesis polymorphism has long served as a model for assessing the climatic drivers of
adaptation in natural populations. Pioneering work in the 1950's and 1960's identified cold

447	winter temperatures as key drivers of reduced HCN frequencies at northern latitudes and higher
448	altitudes (Daday 1954a,b, 1958, 1965). More recent work corroborated the finding that colder
449	environments have reduced cyanogenesis (Ganders 1990; Kooyers and Olsen 2012, 2013) and
450	additionally identified aridity as a correlate of HCN frequencies, with more HCN in drier
451	habitats due to selection favoring plants producing cyanogenic glucosides (Ac) (Kooyers and
452	Olsen 2013; Kooyers et al. 2014). Our results are consistent with previous work demonstrating a
453	cost to producing HCN or its metabolic components in frost-prone habitats (Daday 1954a, 1958,
454	1965; Ganders 1990; Kooyers and Olsen 2013; Kooyers et al. 2018): northern cities with lower
455	winter temperatures and greater frost exposure had reduced cyanogenesis than southern cities. In
456	contrast to previous work (Kooyers and Olsen 2013), we did not identify aridity as an important
457	predictor of mean HCN frequencies, possibly because the latitudinal transect sampled here
458	spanned a shallow aridity gradient (annual aridity index range: 0.84 – 1.22) and steeper gradients
459	may be necessary to detect aridity as an important correlate of HCN frequencies (e.g. New
460	Zealand cline, aridity index range: 0.5941–4.8569, Kooyers and Olsen 2013).

461

462 Urban-rural clines in cyanogenesis

Although the repeated appearance of clines in different cities suggests that selection is acting on HCN, population genetic simulations demonstrated that genetic drift can also generate similar patterns (Santangelo et al. 2018a). Thus, the presence of repeated clines in HCN is insufficient evidence for the role of selection to drive adaptation of urban populations. Two current lines of evidence suggest a negligible role of genetic drift in driving urban HCN clines. First, observed clines in HCN are substantially stronger than those expected under realistic gradients in the strength of drift alone (Santangelo et al. 2018a), suggesting that other processes are additionally

470 contributing to the presence of clines. Second, recent population genetic analyses show no 471 increased strength of genetic drift in urban white clover populations, and clines are evolving 472 despite substantial gene flow between urban and nonurban populations, consistent with natural 473 selection driving local adaptation of urban populations (Johnson et al. 2018). 474 Two additional lines of evidence presented here solidify the role of selection rather than drift 475 in driving the formation urban-rural HCN clines. First, the presence of repeated clines in the 476 same direction at individual loci underlying HCN strongly implicates selection since genetic drift 477 is expected to drive random fluctuations in allele frequencies at a single locus (Santangelo et al. 478 2018a). Second, the negative relationship between the strength of clines and regional winter 479 conditions suggests that latitudinal variation in winter temperature and snow depth-or 480 something correlated with it—modulates the strength of selection along urban-rural gradients, 481 driving phenotypic clines in cyanogenesis. Additionally, the presence of clines at each individual 482 locus, and equivalent deletion haplotype diversity in urban and rural populations across multiple 483 cities, both suggest that selection favors acyanogenic plants rather than alternative biological 484 functions of individual loci or because of genes tightly linked to particular deletion haplotypes. 485 Environmental heterogeneity among 'replicate' environments can reduce the extent of 486 parallel evolution (Stuart et al. 2017). Previous work in white clover identified colder winter 487 temperatures in urban populations as a putative mechanism driving reduced HCN frequencies in 488 urban environments, which was alleviated in cities with high snowfall (Thompson et al. 2016). 489 Based on this earlier work, we predicted the weakest clines in cities with high mean winter 490 temperature or high snowfall due to the absence of frost-mediated selection against HCN in these 491 cities. Consistent with this prediction, cities with high snowfall (i.e. high values along PC1<sub>slope</sub>) 492 had the weakest clines, potentially due to snow buffering plants from frigid temperatures and

493 weakening frost-mediated selection against HCN-producing genotypes. However, the strongest 494 clines occurred in the warmest cities (i.e. low values along  $PC1_{slope}$ ), contrary to our predictions. 495 Importantly, this provides no information about whether lower winter temperatures in cities is an 496 important selective agent; urban frost may still be important in frost-prone cities with shallower 497 urban-rural gradients in snow depth, as suggested by currently available data (Thompson et al. 498 2016). However, this does suggest that alternative mechanisms must drive the evolution of clines 499 in warmer cities where frost is uncommon. Indeed Atlanta, which gets little snow (mean snowfall 500 = 0.07 cm/year) and is relatively warm throughout the winter months (average minimum winter 501 temperature = -0.43 °C) contained the strongest phenotypic cline in HCN observed in any city to 502 date ( $\beta = 0.36$ , Table 1).

503 The stronger clines in warmer cities suggests that regional temperature modulates the 504 strength of selection along urban-rural clines in some cities. Given that cyanogenesis functions as 505 an antiherbivore defense (Dirzo and Harper 1982; Thompson and Johnson 2016; Santangelo et 506 al. 2018b), some clines may be explained by differential herbivory among urban and rural 507 populations if urbanization reduces herbivore damage (Raupp et al. 2010; Moreira et al. 2019). 508 Although previous experimental work found a negligible role of herbivory as a driver of urban 509 clines in HCN (Thompson et al. 2016), this work was performed in a single northern city 510 (Toronto). Since herbivory often increases with warmer temperatures (Lemoine et al. 2014), the 511 role of herbivory in generating urban-rural clines in HCN may be more important in warmer, 512 southern cities. Additional work quantifying the strength of clover-herbivore interactions, and 513 biotic interactions more generally, in these cities is needed. Alternatively, recent experimental 514 data suggests a cost to producing the metabolic components of HCN in stressful environments, 515 especially for cyanogenic glucosides (Kooyers et al. 2018). If environmental stressors are

516	stronger in cities (e.g. frost, salinity, pollution, etc.), costs involved in the production of the
517	metabolic components of HCN may result in selection against these genes and lower frequencies
518	in urban populations. Consistent with this hypothesis, some urban-rural clines in HCN were
519	mirrored only by clines at Ac, suggesting selection may be acting on this locus due to its greater
520	metabolic costs (Kooyers et al. 2018).

521

# 522 *Conclusions and future directions*

523 We have demonstrated the repeated evolution of urban-rural cyanogenesis clines across eastern 524 North American cities. A major goal for future work in this system entails distinguishing among 525 the targets of selection across replicate clines (i.e., HCN vs. Ac. vs. Li) and disentangling the 526 numerous ecological (e.g., environmental factors) and evolutionary (e.g., selection, drift) drivers 527 of (non)parallel responses of HCN to urbanization. This work will require quantifying a broad 528 array of environmental factors at a finer-scale (e.g. population-level) in cities spanning all 529 continents. White clover is a natural model for understanding how cities drive parallel evolution 530 on a global scale due to its ubiquity across the globe and ease of sampling and phenotyping. Such 531 work would advance our understanding of how cities influence the evolution of populations in 532 our own backyards, and further cement the utility of cities as useful models for understanding the 533 causes and consequences of parallel evolution in nature.

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- 668

# 669 **Tables**

670

671 **Table 1:** Beta coefficients (i.e. slope) and *P*-values from linear models testing the change in the frequency of HCN, *Ac*, or *Li* with

672 increasing distance (standardized) from the urban center for each of 16 cities. Also shown are the total number of populations and

plants sampled in each city, and whether deletion haplotypes were identified in urban and rural population of that city. Bolded terms

represent linear clines that were significant at P < 0.05. Grey boxes represent cities where we did not quantify the frequency at the

675 genes underlying HCN.

City	# populations	# plants	Haplotype	$\beta_{\rm HCN}$	$\beta_{\rm Ac}$	$\beta_{\rm Li}$
Atlanta	45	654	Y	0.362***	0.263**	0.161 <sup>‡</sup> *
Baltimore	39	584	Y	0.031	0.065	0.031
Boston	44	876	N	0.119*		
Charlotte	40	589	N	$0.070^{\ddagger}$	-0.077	0.003 <sup>‡</sup>
Cincinnati	40	588	N	0.035		
Cleveland	40	594	Y	0.093	0.067	0.019
Detroit	40	593	N	0.052		
Jacksonville	35	500	Y	<b>0.272<sup>‡</sup>**</b>	0.202**	0.292 <sup>‡</sup> *
Montreal	49	969	N	-0.057		
New York	48	946	Y	0.145*	0.204*	0.033
Norfolk	40	585	Y	0.337***	0.358***	0.038
Philadelphia	40	588	N	-0.031		—
Pittsburgh	40	590	N	0.069		
Tampa†	15	215	N	-0.029		
<b>Toronto</b> ?	121	2379	N	0.283***	0.218**	0.271***
Washington, D.C.	45	658	Y	0.175*	0.326**	$0.062^{\ddagger}$

676 Significance of  $\beta$  values: \*P < 0.05; \*\*P < 0.01, \*\*\*P < 0.001

677 ‡ Cities were better fit by a quadratic model (see online supplementary text: "Assessing the fit on non-linear clines") and showed a significant non-linear change

678 in the frequency of HCN, Ac, or Li with increasing distance from the urban center.

679 † Tampa was excluded from the analysis testing the environmental predictors of the strength of clines since it was functionally fixed for HCN (Fig. 1).

680 INumber of populations and plants for Toronto reflects the total across three urban-rural transects. The coefficients and *P*-values here are from a model that

681 includes all populations along all three transects since all three transects showed significant clinal variation when analyzed independently (Thompson *et al.* 2016)

#### **Figure legends** 682

683

684 Figure 1: Map of 16 cities from which we sampled white clover populations along urban-rural 685 transects. Pie charts represent the mean frequency of HCN (black = HCN+, white = HCN-) for each city when averaged across all populations along the transect. Map color depicts the gradient 686 687 in the minimum winter temperature (MWT, °C) taken from BioClim.

688

689 Figure 2: Urban-rural clines in the frequency of HCN within populations of *Trifolium repens* 

690 across 16 cities in eastern North America. The frequency of HCN within T. repens population is

691 plotted against the standardized distance from the urban center. Solid lines represent linear

692 regressions from cities where the phenotypic cline in HCN was significant at P < 0.05, whereas

693 dashed lines are cities that lack significant clinal variation. The thick black line represents the

694 main effect of standardized distance on HCN frequencies, averaged across all cities.

695

696 Figure 3: Mean HCN frequency was influenced by (a) the number of days below 0 °C with no

snow cover — a measure of frost exposure — and (b)  $PC1_{HCN}$ , a component axis accounting for 697

698 90.2 % of the variation in maximum summer temperature (°C, Bio5), minimum winter

699 temperature (°C, Bio6), annual potential evapotranspiration (mm), monthly summer precipitation

700 (mm), and snowfall (cm) (inset in (b)). City labels are slightly jittered to avoid overlap, if

701 necessary. Cities with low values along PC1<sub>HCN</sub> have high summer temperatures, high minimum

702 winter temperatures, high summer precipitation and potential evapotranspiration, and low

703 snowfall, whereas cities with high values along  $PC1_{HCN}$  have the opposite. (City abbreviations: 704 Jacksonville (Jax); Tampa (Tpa); Atlanta (Atl); Norfolk (Nor); Charlotte (Clt); Toronto (Tor);

705

Montréal (Mtl); Detroit (Det); Washington D.C. (DC); Cleveland (Clv); New York (NY); 706 Pittsburgh (Pgh); Boston (Bos); Baltimore (Blt); Cincinnati (Cin); Philadelphia (Phl)).

707

Figure 4: The strength of urban-rural clines in HCN was influenced by PC1<sub>Slope</sub>, a composite 708

709 axis that accounts for 92.8% of the variation in minimum winter temperature (°C, bio6),

710 maximum summer temperature (°C, bio5), snowfall (cm), and snow depth (cm). City labels are

711 slightly jittered to avoid overlap, if necessary. Bolded cities shower significant linear changes in

712 HCN along urbanization gradients. Cities with low values along PC1<sub>Slope</sub> have little snow and

713 higher minimum winter and maximum summer temperatures, whereas cities with high values

714 along PC1<sub>Slope</sub> have the opposite. City abbreviations are the same as in Fig. 3.







