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Greater aridity increases the magnitude of urban nighttime vegetation-derived air cooling

To cite this article: Peter C Ibsen *et al* 2021 *Environ. Res. Lett.* **16** 034011

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LETTER

Greater aridity increases the magnitude of urban nighttime
vegetation-derived air cooling

OPEN ACCESS

RECEIVED
14 October 2020REVISED
21 January 2021ACCEPTED FOR PUBLICATION
25 January 2021PUBLISHED
15 February 2021

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Keywords: urban heat, vegetation cooling, ecosystem service, aridity, air temperature

Supplementary material for this article is available [online](#)

Abstract

High nighttime urban air temperatures increase health risks and economic vulnerability of people globally. While recent studies have highlighted nighttime heat mitigation effects of urban vegetation, the magnitude and variability of vegetation-derived urban nighttime cooling differs greatly among cities. We hypothesize that urban vegetation-derived nighttime air cooling is driven by vegetation density whose effect is regulated by aridity through increasing transpiration. We test this hypothesis by deploying microclimate sensors across eight United States cities and investigating relationships of nighttime air temperature and urban vegetation throughout a summer season. Urban vegetation decreased nighttime air temperature in all cities. Vegetation cooling magnitudes increased as a function of aridity, resulting in the lowest cooling magnitude of 1.4 °C in the most humid city, Miami, FL, and 5.6 °C in the most arid city, Las Vegas, NV. Consistent with the differences among cities, the cooling effect increased during heat waves in all cities. For cities that experience a summer monsoon, Phoenix and Tucson, AZ, the cooling magnitude was larger during the more arid pre-monsoon season than during the more humid monsoon period. Our results place the large differences among previous measurements of vegetation nighttime urban cooling into a coherent physiological framework dependent on plant transpiration. This work informs urban heat risk planning by providing a framework for using urban vegetation as an environmental justice tool and can help identify where and when urban vegetation has the largest effect on mitigating nighttime temperatures.

1. Introduction

Average nighttime air temperatures (T_{air}) have been steadily increasing across the US in recent years, with implications for reduced human health and well-being [1, 2]. Rising nighttime T_{air} is especially important for heat vulnerability during summer months when heat waves intensify heat-related health effects, and are also a major global cause of weather-related mortality [3–5]. Urbanization exacerbates T_{air}

increases through the urban heat island (UHI) effect, which describes how nighttime T_{air} in cities increases relative to non-urban locations through increased daytime heat storage [6]. With increasing urbanization, more people are becoming vulnerable to inequitable distributions of heat risk and increased cooling costs [7, 8]. Within cities, developed areas with little vegetation result in hot spots where local temperatures are substantially greater than the city-wide average, exacerbating UHI effects [9–11]. Not

only does urban warming elevate health risks, high temperatures also increase energy costs due to air conditioning [8, 12]. While a variety of aspects of urban form can influence urban temperatures such as cool pavements, green roofs, and urban wetlands [13, 14], the cooling capacity of urban vegetation may be a key factor determining the health and well-being of a neighborhood [15]. Increasing urban vegetation density has been proposed as an adaptation approach for mitigating urban nighttime temperatures and resulting heat-related vulnerabilities [16–18]. However, recent work has revealed substantial variability in the magnitude of vegetation-derived nighttime cooling (Delta T_{veg}). In Madison, WI, recent work showed only 1 °C air cooling associated with urban vegetation, while studies in Los Angeles and Palm Springs, CA, reported a mean 2.5 °C and 4.9 °C Delta T_{veg} , respectively [9, 17, 19]. In Salt Lake City, UT, nighttime Delta T_{veg} in vegetated parks reached up to 3.3 °C [20]. A more comprehensive evaluation of Delta T_{veg} is needed, both within and among cities.

Variability in vegetation-derived cooling may reflect multiple physiological mechanisms responsible for air-cooling effects. Urban cooling can occur through shading and by transpiration, both of which alter the urban energy balance by moderating sensible and latent heat fluxes [21–23]. While atmospheric factors such as irradiance intensity and windspeed also influence latent and sensible heat flux [24–26], the positive effect of atmospheric aridity on transpiration indicates that humidity and temperature may influence energy dynamics [27, 28]. Atmospheric aridity can be measured as the vapor pressure deficit (VPD), which describes the difference between the amount of water pressure the atmosphere can hold (saturation vapor pressure) and the total amount of water pressure at a specific temperature (actual water vapor pressure) [29]. Transpiration reduces heat by increasing latent heat flux as energy is used to evaporate water from leaves [6]. While vegetation rooted in dry soils can limit water loss by closing leaf stomata, transpiration from urban vegetation can be maintained even in high heat and aridity because regular irrigation mitigates physiological responses to water limitation [27, 30]. In both urban and natural settings, the tight coupling of VPD and plant transpiration has been well established [31, 32], as are the positive correlations between VPD and Delta T_{veg} within certain cities [33, 34]. However, the direct linking of VPD at the air cooling potential of urban vegetation has not been shown across multiple cities of varying climates and mean summer VPD values.

While Delta T_{veg} exhibits a relationship with VPD, the high variation among results of prior studies investigating urban Delta T_{veg} in different climate regions reveals a gap in our understanding of key drivers for the cooling capacity of urban vegetation [9, 17]. To reconcile differences among the estimated Delta T_{veg} , we ask, how does the magnitude of

vegetation-derived cooling vary within and among cities of the United States? We hypothesize that Delta T_{veg} is predominantly affected by VPD and other atmospheric processes that influence vegetation transpiration.

We test this transpiration hypothesis by measuring the variation of nighttime T_{air} across a vegetation gradient within and among cities representing a range of aridity. By using a novel network of microclimate sensors in eight U.S. cities, we predict that daytime VPD, windspeed, and solar irradiance will be positively correlated with Delta T_{veg} , with daytime VPD having the strongest effect. Among cities, we anticipate that Delta T_{veg} will be greater in cities with higher VPD, and within cities we predict great Delta T_{veg} during weather patterns that increase aridity. The high temporal resolution, spatial extent, and seasonal extent of the sensor networks provided opportunities to observe heat wave events in each city and temporal shifts in local climates such as heat waves and monsoons, which allowed multiple tests of predictions on the importance of atmospheric drivers of transpiration influencing the magnitude of Delta T_{veg} .

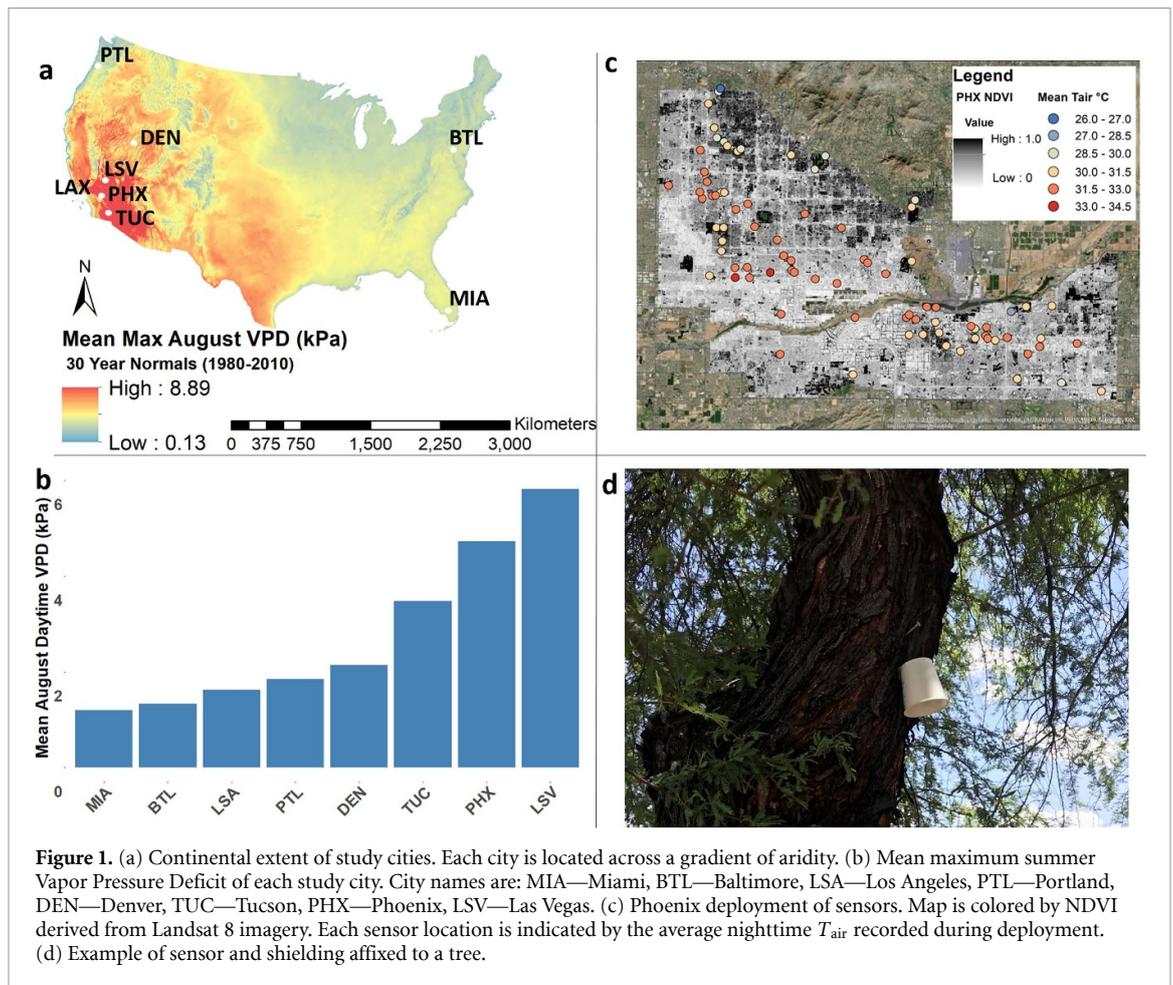
2. Methods

2.1. Study sites

Our study took place within eight major U.S. cities that span a gradient of aridity from Miami (mean August VPD: 0.67 kPa) to Las Vegas (mean August VPD: 5.86 kPa) (figure 1). Each study city was selected based on metro area populations larger than 1,000,000 people and providing a representation of a continental aridity distributions. Each city exhibited a wide range of vegetation density from urban cores to lush parklands (table 1). Sensor distribution within each city was stratified across a gradient of urban vegetation in each city, which was measured spectrally using the normalized difference vegetation index (NDVI). NDVI is a widely used metric of vegetation and encompasses all photosynthetically active vegetation such as trees, grasses, and shrubs [35]. Sensors were deployed in the core of each metropolitan area and the surrounding environs. The extent of deployment differed slightly for each city. All climate data were derived from the nearest airport weather station through the National Centers for Environmental Information (www.ncdc.noaa.gov/cdo-web/datatools/lcd). Across cities, midday VPD ranged from 0.0 kPa to 9.4 kPa, and nighttime cooling magnitude ranged from -0.6 °C/NDVI to 8.3 °C/NDVI.

2.2. Data acquisition and sensor deployment

To test the relationship of vegetation cover, T_{air} , and VPD, we deployed 100 microclimate sensors that logged data at 60 min intervals (Maxim Integrated Products, Inc. iButton Thermocron DS1921 & DS1922L) in each city (100 sensors/city, total 800 sensors). This style of sensor has a history of



being used for rural and urban microclimate analyses [9, 36, 37]. Each sensor was shielded from direct solar radiation in the manner of Crum *et al* (2017), encased in a breathable mesh and housed in custom polystyrene cylindrical white cups [17]. While the precision ($\pm 1^\circ\text{C}$) of iButtons is coarser than a more commonly used instruments such as Campbell Scientific HMP60-L ($\pm 0.6^\circ\text{C}$), to assess the potential discrepancy in our sensors, we validated our sensors for bias and sensitivity against an HMP60-L sensor (see supplementary appendix for more detail). The low cost of iButtons allowed for a large spatial distribution within and among cities throughout the United States to create a continental scale network of city-scale networks collecting *in-situ* data.

We determined values of NDVI with Landsat 8 imagery of each study city, retrieved for cloud-free days during the study period. Using NDVI as our metric of urban vegetation allowed for the required broader comparisons of within and among city cooling, as NDVI quantifies the variety of vegetation found at the continental scale. NDVI was calculated with Landsat 8 bands 4 (Near Infrared) and 5 (Red), using the equation $\text{NDVI} = (\text{Band 5} - \text{Band 4}) / (\text{Band 5} + \text{Band 4})$ in the Raster Calculator function of ESRI ArcGIS 10.6.1. We used the aggregate function of the Spatial Analyst extension in ESRI ArcMap to

scale the native Landsat 8 resolution of 30 m per pixel to a coarser 90 m per pixel, which has been identified as an appropriate spatial scale to observe the signal of vegetation induced cooling [19, 38]. Sensors recorded T_{air} and relative humidity every hour of every night during summer months in one year (i.e. \sim June through September, varying slightly for each deployment) (table 1). The exact time of deployment varied by city, but was between early to late June with sensors recovered in late August to early September (table 1). Locations of sensor deployment were determined by randomly selecting 20 locations within five binned categories of NDVI values spanning the range of NDVI within each city. Binning the distribution of sensors allowed us to capture the full NDVI gradient within each city while also randomizing the sensor placement. Random selection of sites was conducted using the ArcGIS extension Sampling Tool 10 [39], and potential deployment locations were derived as the global positioning system coordinates in the center of a single 90 m pixel [17, 19]. Deployments occurred in 2017, 2018, and 2019. Deployments in each city took approximately 3–5 d, where potential sites for each sensor were located and sensors were affixed to the nearest tree with a full canopy. Sensors were affixed at ~ 2 m height from the surface, and the location was recorded. For areas

Table 1. Metadata for each study city. Including population, Koppen Climate Classification, mean summer VPD, range of NDVI (max pixel NDVI – min pixel NDVI) for city extent, the area of each city covered by our recovered sensors, the number of sensors recovered from initial deployment of 100 and dates of sensor deployment.

City	Population (millions of people)	Climate (Koppen)	Mean max august VPD (kPa)	NDVI range (NDVImax – NDVImin)	Sensor deployment area (km ²)	Sensors recovered (out of 100)	Dates of deployment
Baltimore, MD	2.7	Humid subtropical (Cfa)	1.66	0.788	173.48	78	11 July 2017–30 September 2017
Denver, CO	2.9	Humid continental hot summers with year around precipitation (Dfa)/cold semi-arid climate (BSk)	2.94	0.685	322.42	68	10 July 2018–12 September 2018
Las Vegas, NV	2.2	Hot desert (BWh)	6.32	0.732	360.36	81	11 June 2018–19 August 2018
Los Angeles, CA	13.1	Warm summer mediterranean (CSb)	1.76	0.558	167.90	89	23 June 2017–14 September 2017
Miami, FL	6.1	Tropical monsoon (Am)	1.49	0.657	368.21	80	01 July 18–17 September 2018
Phoenix, AZ	4.8	Hot desert (BWh)	5.71	0.582	277.10	83	15 June 2017–15 August 2017
Portland, OR	2.4	Warm-summer mediterranean climate (Csb)	2.69	0.892	194.76	95	20 June 2017–24 August 2017
Tucson, AR	1.0	Hot semi-arid (BSH)	4.6	0.737	237.37	83	14 June 2019–07 September 2019

in which no suitable tree location could be found near the randomized point, the next nearest location with a similar vegetation density was used. To reduce shading and reflected radiation from nearby structures, trees located next to buildings were not used as sensor sites. Across all cities, ~18% of sensors were lost due to equipment error and/or vandalization.

2.3. Data analysis

Data were downloaded from each sensor and restricted to measurements recorded at 01:00 local time, which was approximately five hours past sunset when UHI effects are estimated to be strongest [6]. To assess vegetation's influence on T_{air} , we first spatially detrended T_{air} patterns within each city of the study. This procedure was needed to remove temperature variation associated with geographic factors such as maritime effects in Los Angeles and Baltimore that were unrelated with vegetation cooling. For each city, the detrending process isolated residuals from a linear trend-surface regression of the NDVI at sensor location against the sensors recording over the entire city ($T_{\text{air}} \sim \text{Lat}_{\text{cityA}} + \text{Lon}_{\text{cityA}}$). The residuals from the regression are the T_{air} values with regional spatial drivers of T_{air} removed and were used as detrended sensor values of T_{air} for subsequent analyses. Vegetation-derived cooling magnitude was then determined by linear regression

analysis, regressing citywide NDVI against detrended sensor T_{air} at 01:00 local time. Significant regression slopes were used as a metric of the vegetation cooling magnitude ($^{\circ}\text{C}/\text{NDVI}$) and used for final analysis [9, 17]. The regression slope is a singular point of citywide Delta T_{veg} for each day of study deployment. The remaining data analysis and presentation uses the significant slope of the regression of NDVI on T_{air} as the primary metric of Delta T_{veg} .

Heat waves were defined as five continuous days where the daily mean maximum temperature is 5°C above the normal mean maximum temperature, calculated from 30 year normals derived from PRISM climate data (PRISM Climate Group, Oregon State University, <http://prism.oregonstate.edu>, created 27 September 2019) [40]. To determine the significance between heat wave and post-heat wave cooling, we conducted a bootstrap randomization procedure where each cooling magnitude was shuffled and resampled with replacement. We ran 1000 variations of resampling, thereby building a data set of heat wave cooling magnitude and post-wave cooling magnitudes. Monsoon season was determined using the pre-2009 National Weather Service definition, which begins an Arizona monsoon season following three continuous days averaging a dew point $> 12.2^{\circ}\text{C}$ (<https://climas.arizona.edu/rgbo/rio-grande-bravo-outlook-june-2017/monsoon-2017>).

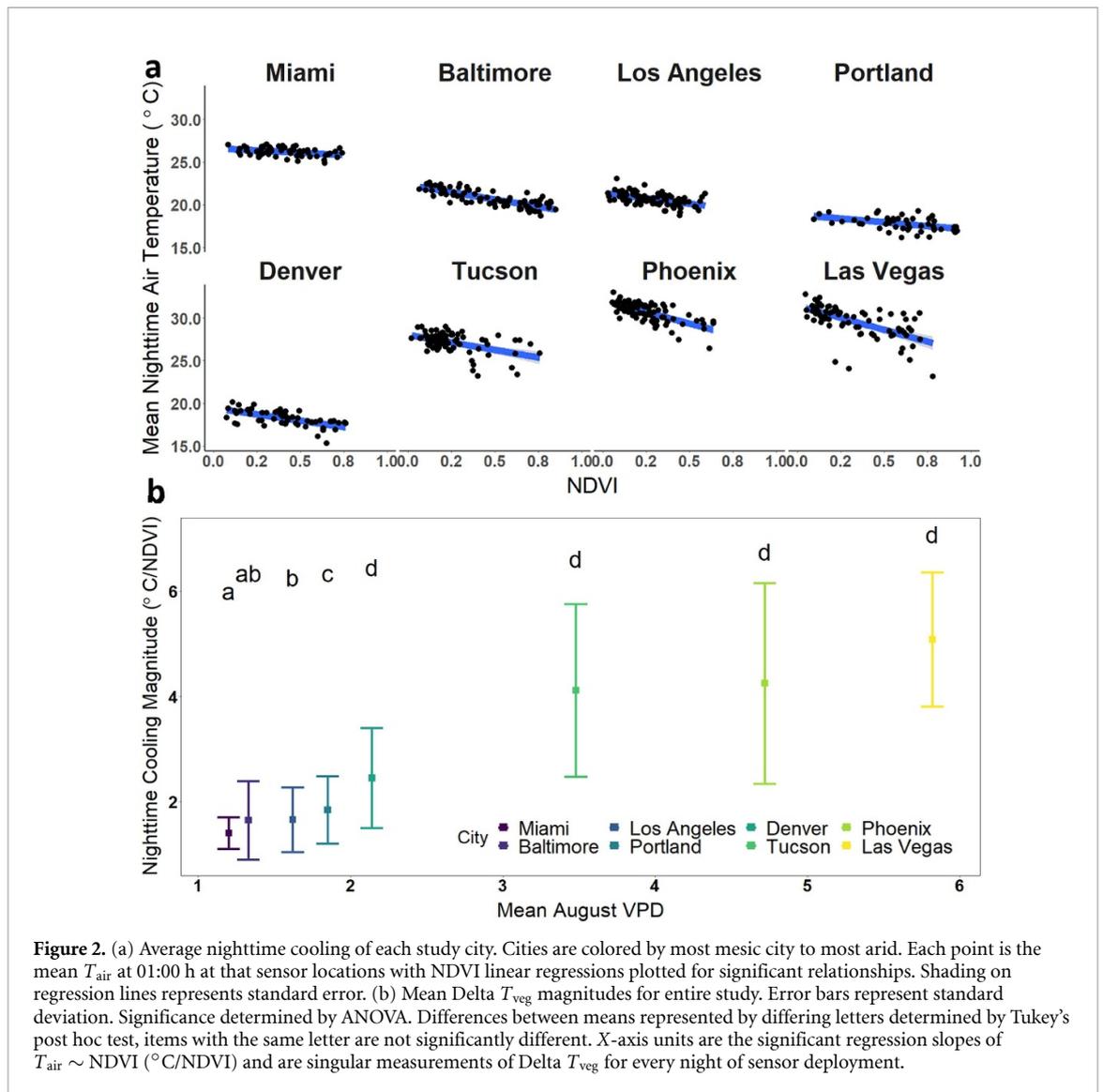


Figure 2. (a) Average nighttime cooling of each study city. Cities are colored by most mesic city to most arid. Each point is the mean T_{air} at 01:00 h at that sensor locations with NDVI linear regressions plotted for significant relationships. Shading on regression lines represents standard error. (b) Mean Delta T_{veg} magnitudes for entire study. Error bars represent standard deviation. Significance determined by ANOVA. Differences between means represented by differing letters determined by Tukey's post hoc test, items with the same letter are not significantly different. X-axis units are the significant regression slopes of $T_{\text{air}} \sim \text{NDVI}$ ($^{\circ}\text{C}/\text{NDVI}$) and are singular measurements of Delta T_{veg} for every night of sensor deployment.

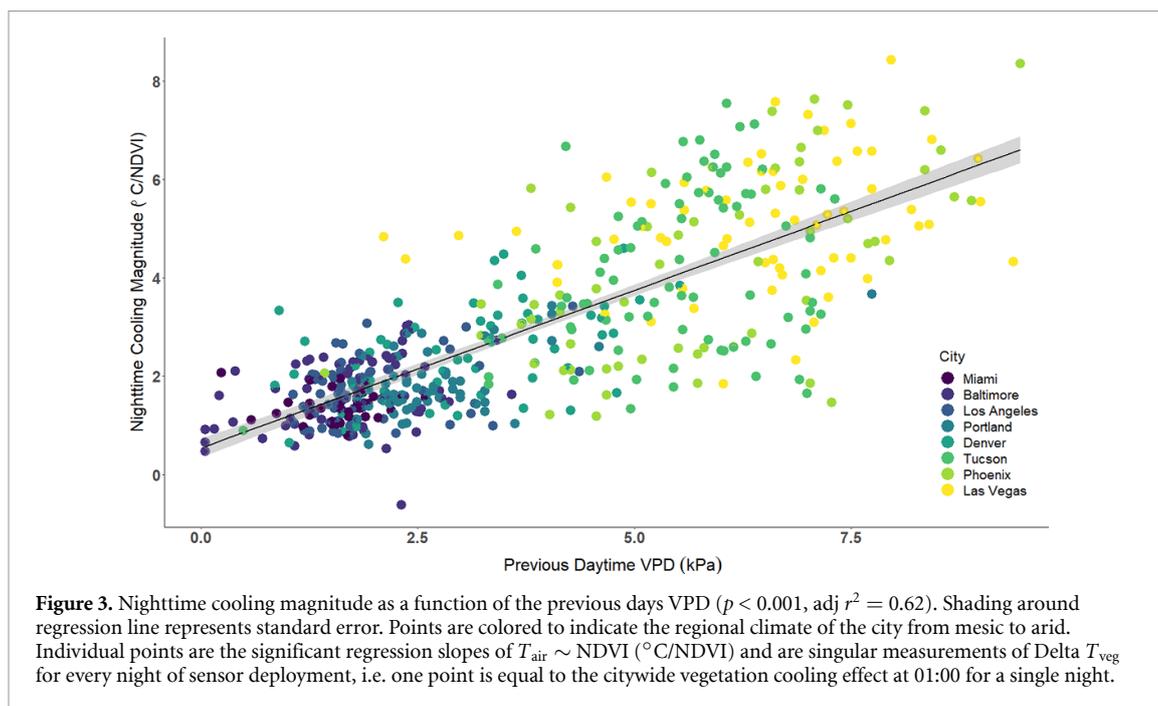
Differences between resulting slopes of the regression of site NDVI and heat-wave T_{air} of each city were determined with an analysis of variance (ANOVA) and subsequent pair-wise city differences were determined with a Tukey's post-hoc test. This method was repeated for analyzing heat waves and monsoonal effects. All statistical analyses were performed in R version 3.60 (R Core Team, 2019).

3. Results

Within all study cities, increased NDVI was significantly correlated with reduced nighttime T_{air} (cities ordered from least arid to most: Miami: $p = 0.002$, $r^2 = 0.14$, Baltimore: $p < 0.001$, $r^2 = 0.63$, Los Angeles: $p < 0.001$, $r^2 = 0.27$, Portland: $p < 0.001$, $r^2 = 0.22$, Denver: $p < 0.001$, $r^2 = 0.39$, Tucson: $p < 0.001$, $r^2 = 0.23$, Phoenix: $p < 0.001$, $r^2 = 0.48$, Las Vegas: $p < 0.001$, $r^2 = 0.40$) (figure 2(a)). We computed Delta T_{veg} magnitudes as the slope of T_{air} and NDVI relationship, which generates a standardized metric of the change in air temperature

across a one-unit change in NDVI (i.e. bare ground to full vegetation). Among cities, the mean cooling magnitude varied, exhibiting a $4.5^{\circ}\text{C}/\text{NDVI}$ range (mean slope and standard deviation of Delta T_{veg} for each city ordered from least arid to most: Miami: mean = 1.41, sd = 0.30, Baltimore: mean = 1.64, sd = 0.74, Los Angeles: mean = 1.65, sd = 0.61, Portland: mean = 1.85, sd = 0.64, Denver: mean = 2.45, sd = 0.95, Tucson: mean = 4.12, sd = 1.64, Phoenix: mean = 4.25, sd = 1.91, Las Vegas: mean = 5.08, sd = 1.27) (figure 2(b)). When comparing citywide Delta T_{veg} across the mean August VPD of each city, the mean effect of vegetation on cooling was significantly greater ($p < 0.05$) in the more arid cities (i.e. Denver, Tucson, Phoenix, Las Vegas) compared to the mesic cities (i.e. Miami, Baltimore, Los Angeles, and Portland).

When comparing Delta T_{veg} to the mean mid-day VPD, calculated using dry bulb T_{air} , relative humidity, and air pressure at the local 13:00 h at the nearest airport weather station, we found significant correlation with the magnitude of



vegetation cooling the following night ($p < 0.001$, $\text{adj } r^2 = 0.62$) (figure 3). Yet, within each city, the midday VPD effect varied with no relationship observed in Miami or Tucson ($p = 0.965$ and $p = 0.237$, respectively) (supplementary figure 1 (available online at stacks.iop.org/ERL/16/034011/mmedia)). Increases in ambient daytime T_{air} also increased ΔT_{veg} across regions, and this effect was smaller than that of VPD alone ($p < 0.001$, $\text{adj } r^2 = 0.47$). Daytime wind-speed had no significant relationship with ΔT_{veg} among cities, but nighttime windspeed correlated with greater air cooling; however, model fit was low ($p < 0.001$, $\text{adj } r^2 = 0.03$). Direct solar irradiance also had a significant positive correlation with ΔT_{veg} magnitudes and model fit was also low ($p < 0.001$, $\text{adj } r^2 = 0.08$) (supplementary figure 2). A multiple linear regression of possible parameters, including VPD, direct irradiance, and nighttime windspeed had a slightly higher model fit than VPD alone ($\text{adj } r^2 = 0.62$).

Temporally, the regional weather patterns differed from the mean atmospheric conditions during the study in two major ways: heat waves and seasonal monsoons. During the study, all cities experienced at least one heat wave in which the daily mean T_{air} was 5°C above the summer average for four consecutive days. Compared with the average cooling magnitude in days following heat waves, ΔT_{veg} magnitude increased in all cities during heat waves. ΔT_{veg} increases ranged from 0.8°C in Miami to 3.9°C in Phoenix. ΔT_{veg} increases during heat waves also significantly scaled with the citywide mean VPD during heat waves within each city ($p = 0.035$, $\text{adj } r^2 = 0.46$) (figure 4(a)). Two cities (Phoenix and Tucson) experienced a shift from a dry pre-monsoon climate to a more humid monsoonal

climate. Ambient T_{air} and VPD decreased during Phoenix's monsoon season (T_{air} : $p < 0.001$, VPD: $p < 0.001$). In Tucson, ambient T_{air} during the monsoon was consistent with pre-monsoon temperatures ($p = 0.3$), but VPD decreased from pre-monsoon to monsoon season ($p = 0.005$). Transitioning into the monsoon, the magnitude of vegetation cooling in Phoenix decreased from a mean of 8.1°C to 4.6°C ($p < 0.001$), whereas in Tucson, the magnitude of ΔT_{veg} decreased from a mean of 5.0°C to 4.0°C ($p < 0.01$) (supplementary figure 3).

4. Discussion

Increases in urban vegetation cover decrease nighttime T_{air} in cities throughout the United States; nevertheless, the magnitude of cooling varies dramatically among and within cities. Furthermore, not only was higher aridity during heat waves associated with greater cooling, but a consistent pattern also occurred in the two cities that experienced a shift between a hot and dry pre-monsoon and a more humid monsoon weather pattern. The intra-urban temporal changes in ΔT_{veg} are consistent with the patterns observed among cities with respect to the role of daytime transpiration in nighttime cooling. Our results, spanning spatial scales of intra-urban to continental and temporal scales from individual evenings to the summer season, suggest widespread urban vegetation cooling that is consistently sensitive to aridity, implying greater latent heat fluxes in more arid cities at the same levels of vegetation coverage. The strong correlation between aridity (VPD) and the magnitude of ΔT_{veg} is consistent with the hypothesis of transpiration as a primary influence on cooling.

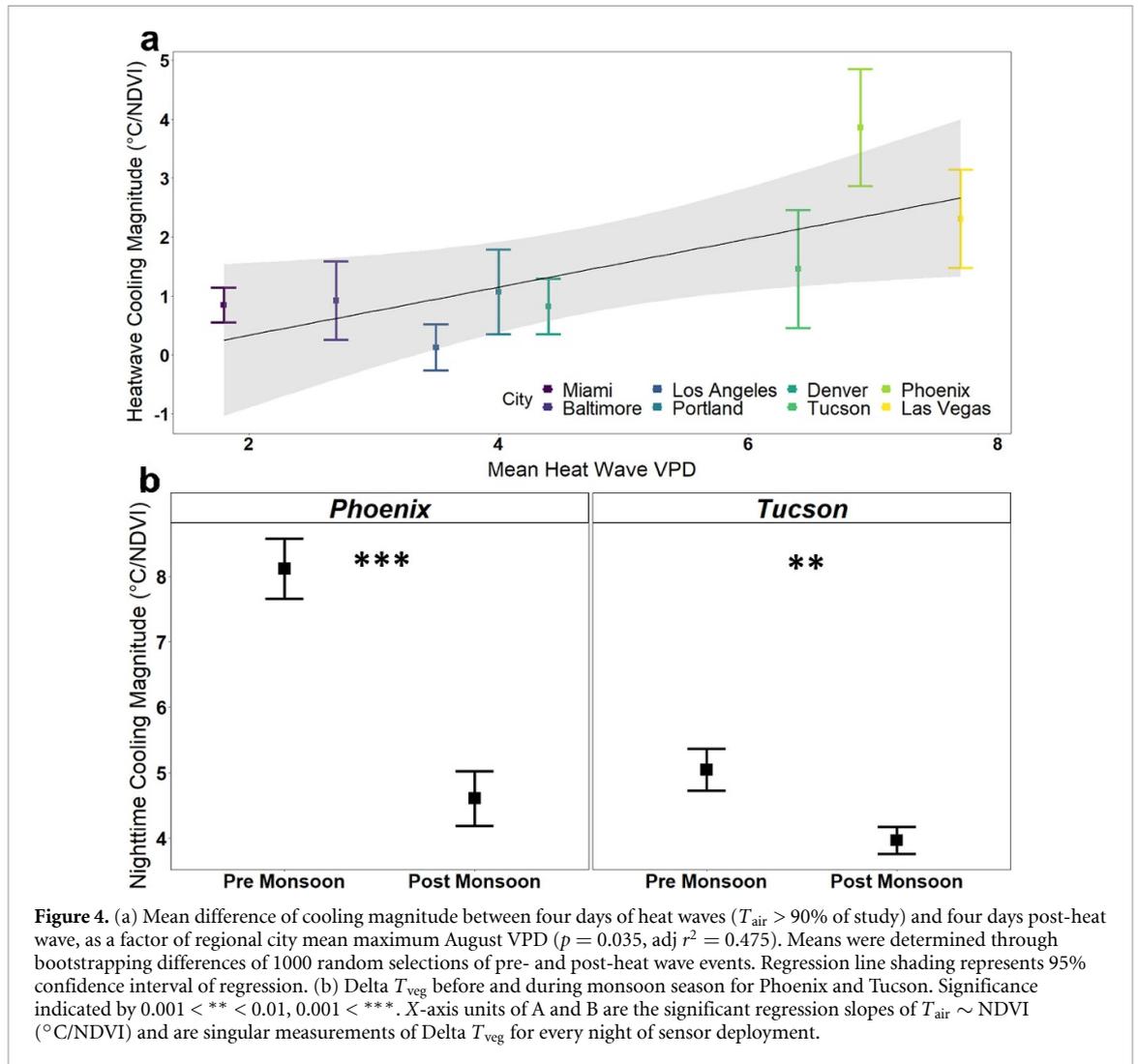


Figure 4. (a) Mean difference of cooling magnitude between four days of heat waves ($T_{\text{air}} > 90\%$ of study) and four days post-heat wave, as a factor of regional city mean maximum August VPD ($p = 0.035$, $\text{adj } r^2 = 0.475$). Means were determined through bootstrapping differences of 1000 random selections of pre- and post-heat wave events. Regression line shading represents 95% confidence interval of regression. (b) Delta T_{veg} before and during monsoon season for Phoenix and Tucson. Significance indicated by $0.001 < ** < 0.01$, $0.001 < ***$. X-axis units of A and B are the significant regression slopes of $T_{\text{air}} \sim \text{NDVI}$ ($^{\circ}\text{C}/\text{NDVI}$) and are singular measurements of Delta T_{veg} for every night of sensor deployment.

The dependence of Delta T_{veg} on VPD helps reconcile the large variation observed in the magnitude of vegetation cooling observed in modeling studies, surface temperature observations, and T_{air} measurements within individual regions. Our results provide needed validation that explains differences in vegetation cooling among cities using regional climate [41, 42] and microclimate [43] models that show varying cooling effects among cities and within neighborhoods. Similarly, our results indicating a strong influence of VPD on Delta T_{veg} are consistent with evaluations of satellite-derived land surface temperatures [16, 44], which also show increased daytime surface cooling in more arid conditions. Importantly, these findings suggest the large variation in observed vegetation cooling effects on urban nighttime T_{air} in varying climates may reflect a consistent underlying cause. Recent work quantifying T_{air} over vegetation gradients within individual cities reported a mean Delta T_{veg} magnitude of 5.9 $^{\circ}\text{C}$ in Palm Springs, CA [17], 3.1 $^{\circ}\text{C}$ in Los Angeles, CA [9], and a Delta T_{veg} range of 0.5 $^{\circ}\text{C}$ –1.1 $^{\circ}\text{C}$ in Madison, WI [19]. When factoring the mean summertime VPD of Palm Springs, Los Angeles, and Madison into our

results (5.4 kPa, 1.8 kPa, and 1.28 kPa respectively), we find these cooling magnitudes consistent with the results presented here that indicate a linear trend of Delta T_{veg} in response to aridity. This linear relationship between VPD with Delta T_{veg} provides a coherent transpiration-based hypothesis for extending studies within individual regions, satellite observations, and modeling approaches.

4.1. Atmospheric drivers of vegetation-derived nighttime cooling

The transpiration hypothesis for regulating Delta T_{veg} suggests the importance of physiological interactions between vegetation and the urban environment. Not only is VPD a key driver of transpiration, but it also strongly correlates with Delta T_{veg} as irrigated vegetation may maintain open stomata in high VPD conditions [27]. Vegetation stabilizes surface temperature in high heat and aridity through maintaining transpiration [45]. As daytime VPD increases, urban vegetation increases transpiration, which in addition to increasing immediate latent heat flux, reduces leaf surface temperature and eventual re-radiation of stored heat energy. The combination

of these modifications to the surface energy balance are the potential causes for our observed correlation between daytime VPD and Delta T_{veg} .

As aridity increases, plants in natural areas generally exhibit a saturating relationship between transpiration and ambient VPD [22, 46], where leaves restrict their stomatal conductance in response to atmospheric drought to limit risks of xylem cavitation. However, transpiration for urban trees is less limited by high VPD relative to their rural counterparts [27]. Transpiration rates are also dependent on local soil moisture, which is generally elevated through irrigation in cities [47, 48]. We found a linear relationship between daytime VPD and Delta T_{veg} , which implies that irrigated urban vegetation experiences fewer limitations to transpiration even in conditions with high atmospheric demand. The increased transpiration resulting from irrigated vegetating experiencing high VPD could cause cooling when combined with higher wind speeds, and our results also indicated a significant, though weak, effect of nighttime wind speed on cooling (supplementary figure 2). Wind-induced air cooling is generally caused by a thinning of the leaf boundary layer that increases latent heat flux [49]; however in practice, correlations between instantaneous wind speed and air cooling are weak [25, 50]. At the continental scale, aridity appears to be the primary driver of changes in cooling magnitudes.

In addition to among-city differences in Delta T_{veg} , within-city temporal differences in Delta T_{veg} during changing weather patterns are also consistent with a hypothesized transpiration mechanism. During heat waves, plants may maintain transpiration rates [30], thereby increasing energy lost to the latent heat of evaporation and resulting in potentially greater localized air cooling. Both Phoenix and Tucson experienced greater nighttime cooling before their monsoon periods, but the effect was highest in Phoenix. Phoenix's midday VPD and T_{air} were both significantly higher than during the monsoon. However, while Tucson's daytime VPD was higher pre-monsoon, daytime air temperature was not significantly different (supplementary figure 3), which was a trend consistent with the hypothesis that VPD is a stronger driver of transpiration derived cooling compared to air temperature alone. With continued irrigation, the strength of the cooling-VPD relationship could increase with climate change in many arid regions that are projected to become more arid [51].

4.2. Using urban vegetation to reduce heat-related health risks

Our study highlights the potential benefits of urban vegetation for mitigating extreme urban heat scale across cities of dramatically different climates. The linear relationships between vegetation and air temperature in all cities (figure 2(a)) imply that adding vegetation in these cities results in a continuous

cooling effect. During periods of high heat, people require a cooler nighttime temperature to sleep and reduce stress on the body [7, 52]. Reducing summertime nighttime urban heat can reduce heat stress and the need for medical treatment due to heat-related symptoms [52, 53]. Mitigating high nighttime T_{air} is important for all urban residents; however, heat-associated health effects are unevenly distributed. In the U.S., lower income and non-white racial-ethnic groups more often live in areas with less vegetation and more impervious surfaces, which is associated with hotter temperatures [54]. Conversely, individuals with higher incomes more frequently live in areas with greater vegetation cover that they can afford to irrigate, which keeps local microclimates cooler [55, 56]. Households with higher incomes can also more readily often afford air conditioning during extreme heat conditions [57, 58]. These inequalities may be exacerbated in the future as mean nighttime air temperatures are projected to increase even more than daytime temperatures [59]. For economically- and racially-marginalized groups living in areas of minimal urban vegetation, heat risks (e.g. exposure to high nighttime T_{air}) may lead to greater vulnerability to health impacts when no coping methods (e.g. air conditioning or denser vegetation) are available.

Municipal policies considering increasing urban vegetation to mitigate heat vulnerability should also consider key trade-offs in achieving cooling. Our results show a strong influence of aridity on the magnitude of vegetation cooling; thus, in order to maintain Delta T_{veg} as an urban ecosystem service, vegetation needs to maintain transpiration. As transpiration is highly correlated to soil moisture, and irrigation practices in many cities determine soil moisture, maintaining Delta T_{veg} highlights a trade-off between cooling services and urban water use [16, 60]. The magnitude of Delta T_{veg} and expected corresponding transpiration rates are highest in cities where municipal water use concerns are of high priority. Poorly planned and extensive tree planting campaigns in arid cities may exacerbate local water shortages through increasing irrigation costs [61]. The water-for-cooling trade-off does have potential for mitigation through adaptive urban management policies of refocusing irrigation and promoting more water-conservative vegetation. In Phoenix, modeling assessments suggest nighttime air cooling could be increased with only a 2.6% water increase by shifting irrigation from areas of dense vegetation to those with sparse vegetation coverages [62]. The trade-off between air-cooling and urban water use can also be lessened by shifting vegetated landcover from primarily turf to trees and shrubs that are adapted to the local climate. For example, in the arid climate of southern Israel, the cooling efficiency of trees with no surrounding grass is 27.5 times higher than exposed turf [63].

Addressing the cost-benefit of implementing urban heat-risk management is a key goal of urban sustainability solutions [40]. Directed urban greening projects also need to be developed for specific cities. Based on our results, greening a downtown block in Baltimore (\sim NDVI = 0.09) so it resembles more of a tree-lined residential street next to a park (\sim NDVI = 0.46) would reduce nighttime air temperature by approximately 0.60 °C. The same greening program in Phoenix would result in approximately 1.3 °C of cooling. The scaling of cooling is city-specific, but discrete increases in urban greening can result in significant nighttime cooling benefits depending on the aridity of the city. Partitioning the relative cooling potentials of vegetation types beyond the total greenness metric of NDVI, in both arid and mesic cities, is a necessary next step in resolving the nexus of urban vegetation, water use, and cooling benefits. Using NDVI as a greenness metric allowed this study to compare vegetation broadly across multiple cities addressing a gap in the research of studies examining the within and among city variation in vegetation derived cooling. While some research has begun partitioning out the amount of cooling provided through shade and transpiration of urban trees [64], identifying cooling rates for different types of urban vegetated parcels [65], or modeling the potential cooling benefits of lawns and trees in a city [18], a multi-city empirical study of vegetation cooling effects delineated by vegetation type should be the next research gap addressed. Future studies should explore other covarying mechanisms of T_{veg} such as the interactions of urban vegetation, albedo, and other reductions in radiant temperature through shading. Albedo can have a significant negative correlation with nighttime surface temperature [15], and urban trees directly reduce surface temperatures through reducing incoming radiation [66]. Land surface cooling is a key component of urban heat health issues, and while out of the purview of this study of air temperatures it should not be ignored in future urban heat mitigation plants.

Our results emphasize the need for city-specific plans focused on urban vegetation as a tool to mitigate high temperatures and region-specific water limitations. Policies must be city-specific, as we have seen that increasing greening by the same amount in Las Vegas and Miami will result in significantly more cooling benefits and subsequent health benefits in arid Las Vegas. Urban stakeholders across regions can use our results to identify areas that would likely receive the most cooling benefit from increases in vegetation, especially as the negative health effects caused by urban heat are not equally distributed. Many cities have developed large scale urban forestry projects [67, 68], but without directing greening initiatives to certain areas of cities, there may be overall increases in cooling that do not address urban heat inequities. To directly address the inequity of urban

heat, rather than to focus on large city-wide greening efforts, policy makers should consider focusing urban greening to areas with the least amount of existing vegetation while also minimizing gentrification and displacement [69, 70]. As people continue to move to cities, using urban vegetation to reduce inequities in heat exposure can contribute to a more sustainable future.

Data availability statement

The data that support the findings of this study are openly available at the following URL/DOI: http://erams.com/s/nsR7qpFiEqUsBhm2quqZg/Urban_Veg_Cooling.

Acknowledgments

We greatly appreciate the logistical and organizational support from the Urban Water Innovation Network, especially including Mazdak Arabi and Sarah Millonig. We thank Dion Kucera, Julie Ripplinger, Holly Andrews, Mia Rochford, Samuel Meltzer, Ariane Middel, Ales Urban, Erin Everton, and Tess Gallagher for field assistance. We thank Steven Crum for helpful comments on the development of the analysis, and Janet Franklin and Louis Santiago for comments on an earlier draft of this paper. We acknowledge funding from the US National Science Foundation (CBET-1444758 and CNH-1924288).

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