Ambient Air Pollution Exposure among Individuals Experiencing Unsheltered Homelessness

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Introduction

Exposure to ambient air pollution is increasingly recognized as a major driver of morbidity and mortality. Ambient air pollution is anticipated to increase as a result of climate change, extreme weather events and wildfires. Within the United States, disparities already exist in exposure to air pollution. Residing in a non-white majority or low-income census tract is associated with increased exposure to fine particulate matter [PM ≤ 2.5 µm in aero-dynamic diameter (PM_{2.5})]. The pattern of air pollution exposure among other vulnerable populations has yet to be established.

Individuals experiencing homelessness represent a growing population in the United States, with over half a million people homeless in 2020.⁵ Among this group, an increasing proportion experience unsheltered homelessness—defined as residence in the street or in a structure not intended for human habitation.⁶ These individuals are uniquely vulnerable to the impact of worsening air quality, particularly outside of large urban centers, where access to indoor shelters may be limited.⁷

We hypothesize that in addition to global air pollution, individuals experiencing unsheltered homelessness are exposed to excess air pollution as a result of proximity to stationary and mobile sources. By considering this potential exposure at the local level, our aim was to provide a broad estimate of exposure, and develop a framework of local-level geospatial analysis that can be used to guide further targeted research and intervention.

Methods

In collaboration with Tulare County Health and Human Services Agency, individuals experiencing unsheltered homelessness were invited to participate in a local knowledge mapping (LKM) survey. Participants were asked to map places of importance to them on a printed base map, with focus on "safe" spaces where they had spent >1 month over the preceding year. Safety was self-defined by participants, and an explanation provided by the participant for each site. Location data was subsequently digitized and mapped at the point level using ArcMap (version 10.7.1; ESRI). Emitters were categorized by major emitter group [PM, organic gas emissions (total organic gases, TOG)], nitrous oxides (NO_x) and sulfur oxides (SO). At the suggestion of local homeless service providers, participants were compensated with \$5 gift cards for McDonalds or with bus passes. This study was approved by the Case Western Reserve University Institutional

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The authors declare they have nothing to disclose.

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Review Board (No. 20191570). Informed consent was obtained verbally using a provided script.

Stationary source emission exposure was estimated using California Air Resource Board (CARB) stationary source emission data. Mobile source emissions across the 2019 calendar year were estimated using the CARB 2021 EMissions FACtors (EmFAC) modeling system. This system provides a county-level estimate of roadway-associated emission modeled for aggregate levels of road speed and seasonal temperature and traffic variation. Mean emissions of (PM_{2.5}), PM \leq 10 μ m in aerodynamic diameter (PM₁₀), TOG, NO_x, and SO were estimated in tons per year.

Three hundred-meter and 1,000-m buffers were generated around stationary emitters and roadways. Using spatial joining, the number of individuals experiencing homelessness whose day or nighttime locations fell within these buffers was calculated at each geographic level, stratified by emission type and location.

Results

A total of 62 individuals experiencing homelessness participated. The majority of participants (68.9%) had resided in Tulare County for more than 10 y. More than half (59.2%) self-identified as female, with 49.2% of participants falling between 45 and 65 years of age. Participants identified a total of 166 locations where they spent the night during the 2019 calendar year and a further 117 "safe" day locations. Fifty-six percent of "safe" day locations intersected with "safe" night locations.

When analyzed by proximity buffer, 32.5% of night locations and 52.1% of day locations were within 300 m of a major roadway. Mobile emissions modeling for the 2019 calendar year estimated total road traffic-related emissions of $PM_{2.5}$ at 93.9 tons/y, with a further 190.5 tons/y of road traffic-related PM_{10} emissions. Total roadway-associated nitrogen dioxide (PM_{20}) emissions were estimated at 186.3 tons/y, with 1541.9 tons/y of TOG emission and 24.5 tons/y of SO₂ emissions.

In all, 891 registered stationary emitters were identified within Tulare County. Proximity of both day and night locations to stationary emitters was common throughout the county (Table 1).

Discussion

Individuals experiencing chronic unsheltered homelessness are exposed to a range of sources of air pollution, extending beyond the exposures captured by ambient air pollution monitoring. Within our sample, >50% of participants reported spending their daytime hours in close proximity to major roadways. Near roadway proximity has been associated with an increased risk of respiratory and cardiovascular disease across multiple cohorts. Proximity to emission sources was also common. Reliance on stationary ambient monitor data may underestimate both potential airborne pollutant exposure and health impacts associated with exposure to these pollutants for this population.

Local knowledge mapping survey techniques represents a lowcost and easily reproducible mechanism to capture activity patterns among individuals exposed to air pollution. When combined with

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Table 1. Percentage of "safe" day and night locations in a 300- and 1,000-m proximity to stationary emitters across Tulare county by site of point in a time—count magnet event.

		Tulare		Visalia		Porterville	
Primary emitter type and proximity		Day locations $[(n=36) \%]$	Night locations $[(n=87) \%]$	Day locations $[(n=61) \%]$	Night locations $[(n=55) \%]$	Day locations $[(n=20) \%]$	Night locations $[(n=24) \%]$
PMT emitter	<300 m	11.0	21.8	32.8	36.3	35.0	8.3
	<1,000 m	80.5	80.4	95.0	89	85.0	66.6
NO _x emitter	<300 m	16.7	20.6	14.8	21.8	65.0	41.6
	<1,000 m	30.6	37.9	59.0	43.6	100	95.8
SO emitter	<300 m	27.8	29.9	36.0	40.0	40	25.0
	<1,000 m	88.9	89.7	98.3	98.2	90	70.8
TOG emitter	<300 m	58.3	49.4	70.5	60.0	55.0	41.7
	<1,000 m	97.2	95.4	98.3	98.2	100	95.8

Note: NOx, nitrous oxide; PMT, total particulate matter; SO, sulfur oxides; TOG, total organic gases.

local spatial data, this can provide guidance regarding further monitoring and policy interventions at a geographically granular level.

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