

# Police-Recorded Crime and Disparities in Obesity and Blood Pressure Status in Chicago

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**Background**—The purpose of this study was to examine associations between several types of police-recorded crime (violent, nonviolent, and homicide) and cardiometabolic health (obesity and elevated blood pressure [BP]), and to determine if associations were modified by age and sex.

**Methods and Results**—We analyzed cross-sectional data (N=14 799 patients) from 3 primary care clinics at an academic medical center in Chicago, IL. Patient-level health data were obtained from the electronic health record (June 1, 2014–May 31, 2015), including body mass index and BP, and linked to the City of Chicago Police Data Portal. Geocoded crime counts were aggregated to census tract and calculated as the annual crime rate per 1000 population. Generalized linear mixed models were used to assess obesity and BP status as a function of crime rate quartile, controlling for patient, clinic, and neighborhood characteristics. Median violent crime rates in each quartile ranged from 15 to 84 per 1000 population. Median age was 56 years (interquartile range, 38–72 years); 42% of patients were obese and 33% had elevated BP. Compared with patients living in the lowest quartile, patients living in the highest quartile for violent crime had 53% higher adjusted odds of obesity (95% confidence interval, 1.15–2.03) and 25% higher adjusted odds of elevated BP (95% confidence interval, 1.01–1.56). In subanalyses examining homicide, a relatively rare event, exposure was not associated with obesity and was inconsistently associated with elevated BP.

**Conclusions**—In a densely populated, high-poverty region in Chicago, recurrent exposure to high rates of violent crime was consistently associated with obesity and elevated BP, but rare exposure to homicide was not. (*J Am Heart Assoc.* 2018;7:e008030. DOI: 10.1161/JAHA.117.008030.)

**Key Words:** epidemiology • high blood pressure • obesity • risk factors • women

Violent crime has increased in 13 major US cities since 2014, including Los Angeles, CA; Houston, TX; Baltimore, MD; and Chicago, IL.<sup>1</sup> Affected communities have been disproportionately urban, with high levels of concentrated

poverty and residential segregation by race.<sup>1,2</sup> These same communities have a high burden of chronic health conditions, especially obesity and hypertension, raising concern about the potential relationship between exposure to violent crime and risk of preventable and costly cardiometabolic diseases.<sup>3,4</sup>

Prior studies have hypothesized that people living in neighborhoods with high rates of violent crime may experience high levels of psychological distress that result in adverse health behaviors and outcomes.<sup>5,6</sup> For instance, studies of US children have documented lower rates of physical activity and higher rates of obesity in unsafe neighborhoods.<sup>7–9</sup> Sandy et al examined 36 936 US children and found that higher violent crime rates were associated with higher body mass index (BMI).<sup>8</sup> However, studies of adult populations and BMI have been more mixed.<sup>7,10</sup> Lovasi et al examined a sample of 13 102 adult New York City residents and found no relationship between homicide rates and BMI.<sup>11</sup> Other adult studies have documented mixed findings for physical activity and other metabolic health outcomes.<sup>12,13</sup>

One possible explanation for these mixed findings is that many prior studies have assessed neighborhood crime using subjective self-reported measures of neighborhood safety.<sup>7</sup>

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## Clinical Perspective

### What Is New?

- Small area analysis using objective measures revealed a consistent relationship between exposure to violent crime and cardiometabolic health (body mass index and blood pressure).
- Associations between violent crime and cardiometabolic health were more pronounced among women.
- Recurrent exposure to “everyday violent crime” may be more strongly associated with cardiometabolic health than a single exposure to severe violent crime.

### What Are the Clinical Implications?

- Exposure to violent crime may be an important social determinant of health in vulnerable populations.
- In an era of population health, healthcare organizations located in high-crime regions may have incentive to participate in violence prevention and intervention efforts.

However, self-reported measures may be systematically biased by individual characteristics, including how people perceive crime or hear about crime in their neighborhood. Allostatic load theory explains how cumulative prolonged exposure to stressors (eg, frequently hearing gunshots at night) may activate physiologic response pathways that result in metabolic or autonomic dysfunction.<sup>14,15</sup> For instance, excess cortisol secretion, resulting from fear of violent crime, can impair glucose metabolism, promote abdominal fat deposition, and increase vascular reactivity.<sup>14,15</sup> These physiologic processes can occur unconsciously and may be difficult for individuals to quantify by self-report.<sup>16,17</sup> Prior studies have also used self-reported measures of health,<sup>7,10</sup> which may lead to significant underestimation of health effects in communities with lower education (eg, poor health literacy)<sup>18</sup> and unreliable access to health care.<sup>19</sup>

Other studies have used objective measures of crime, but they have not examined the different types of crime encountered (eg, violent crime and homicide) or demographic groups targeted (eg, elderly people and women), limiting comparative inference.<sup>10</sup> Stress responses may fundamentally differ by exposure type, and scholars have theorized that violent crimes may be more closely linked to fear-based physiological processes than nonviolent crimes.<sup>10,20</sup> Similarly, responses may vary on the basis of a person’s perceived risk for being targeted. Prior literature has suggested that women and elderly adults may be more vulnerable to fear of victimization and, often, victimization itself.<sup>7</sup> Weinstein et al examined 12 767 adults and found that perceived safety was more strongly associated with physical activity in women than in men.<sup>21</sup> These differences may be of special importance for

identifying and reducing disparities in women’s cardiovascular health, particularly among vulnerable populations.<sup>22</sup>

More important, linkages between neighborhood crime and hypertension have been understudied. Although several prior studies have used objective crime rates to examine associations with physical activity and obesity,<sup>10</sup> fewer studies have used objective crime rates to examine associations with hypertension. In the most compelling study to date, Kaiser and colleagues used data from the Multi-Ethnic Study of Atherosclerosis and found no relationship between self-reported safety and incident hypertension, after adjusting for race/ethnicity.<sup>23</sup> Although this study used objective blood pressure (BP) measurements to assess incident hypertension, it used a survey-based method to assess neighborhood safety.<sup>23</sup> Few studies, if any, have used objective measures of both crime and BP to examine this relationship.

Building on this prior work, this study uses objective measures to examine associations between several types of police-recorded crime (violent, nonviolent, and homicide) and cardiometabolic health (obesity and elevated BP) in a large clinical sample of adults in Chicago. As a secondary objective, this study also examines if associations are modified by age and sex. We hypothesized that use of objective measures would reveal positive associations between neighborhood violent crime rates and obesity and BP status. On the basis of prior studies,<sup>7,21</sup> we also hypothesized stronger associations among women and elderly adults.

## Methods

This study was conducted in a 324 census tract region on Chicago’s south and west sides, a densely populated urban region with ≈992,000 residents, 58% of whom identified as non-Hispanic black and 30% of whom identified as Hispanic or Latino. Many residents in this region were unemployed (21%) and/or living below the federal poverty level (30%).<sup>24</sup> This study region was within the catchment area of clinical sites included in the parent study, CommunityRx, and was selected because of high rates of poverty and preventable illness in the target population. Our study population included all patients (N=14 799) living in the study region and seen in at least 1 of 3 primary care clinics at an urban academic medical center between June 1, 2014 and May 31, 2015. The only patients excluded from the CommunityRx study population were those who had a home address outside the study region; few were excluded because of missing covariate data (2.5%). The median sample size was 100 patients per census tract (interquartile range, 48–250 patients).

This study was conducted with a waiver of informed consent and approval from the University of Chicago (Chicago, IL) Institutional Review Board.

## Data Sources and Measures

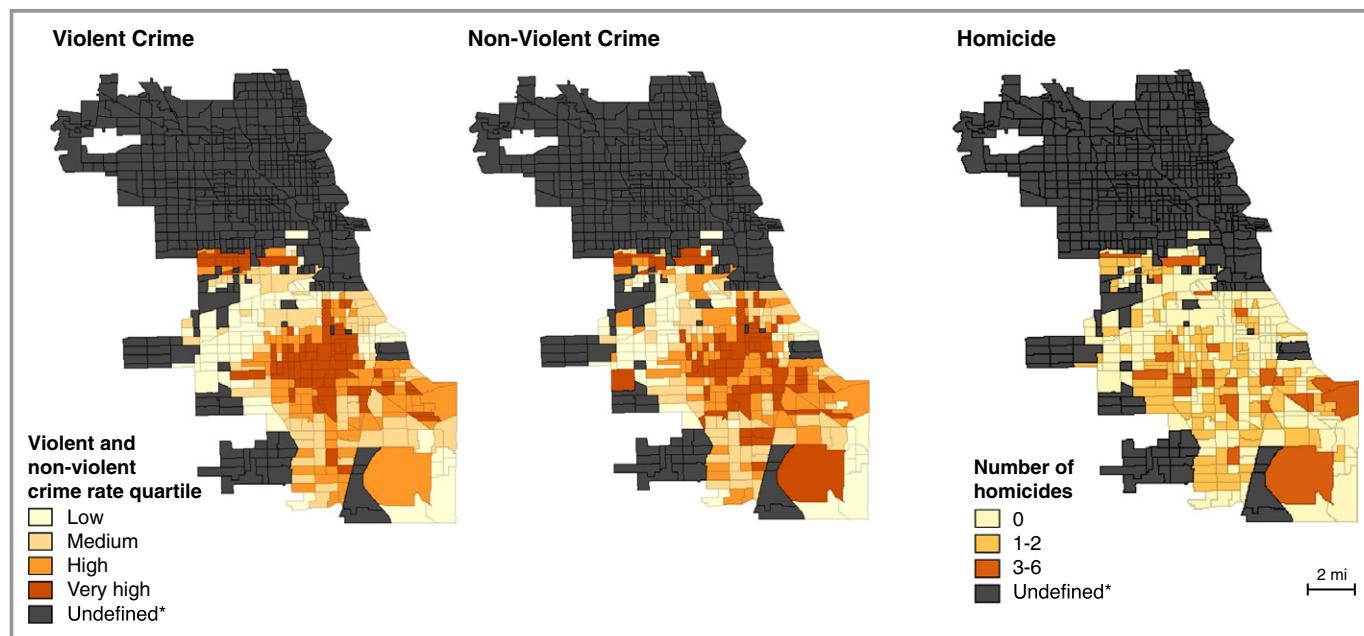
Geocoded crime counts were obtained from the City of Chicago Police Data Portal, a record of every incident of crime reported to or known by the Chicago Police Department. Data were publicly available and extracted weekly from the Chicago Police Department's Citizen Law Enforcement Analysis and Reporting system (<https://data.cityofchicago.org>). These data were linked to contemporaneous patient-level data obtained from the CommunityRx database.<sup>25</sup> The CommunityRx database included clinical information from the electronic health record systems of 3 primary care sites included in this study (internal medicine, obstetrics and gynecology, and geriatrics), including BMI and BP measurements.

The dependent variables were obesity and BP status. Obesity status was defined as  $\text{BMI} \geq 30 \text{ kg/m}^2$ , and elevated BP was defined as systolic BP  $>140 \text{ mm Hg}$  or diastolic BP  $>90 \text{ mm Hg}$ . We chose to analyze BMI and BP as a binary outcome for clinical interpretability. For instance, a higher BP is not necessarily a poorer outcome unless it crosses the previously specified diagnostic threshold. However, we have included sensitivity analyses using continuous values for BMI and systolic BP in Table S1. The CommunityRx intervention, described in previously published work,<sup>25</sup> was implemented during this study period; we thus abstracted the first recorded BMI and BP measured before the intervention to avoid any contamination effects. Standard procedure for the clinics represented in this study included a single BP measurement

at the time of clinical encounter. Thus, those classified as having elevated BP included patients with aforementioned criteria, regardless of diagnosis status, including those who were diagnosed but uncontrolled. Patients who were classified as not having elevated BP included those with well-controlled hypertension.

The independent variable in our analysis included all types of violent crime (assault, battery, criminal sexual assault, robbery, or homicide). We also included nonviolent crime (theft or criminal property crime) as a comparator group, reasoning that nonviolent crime should have a weaker relationship with health than violent crime. Crimes were classified as violent or nonviolent on the basis of the Chicago Police Department Crime Type Categories, derived from the Federal Bureau of Investigation's National Incident-Based Reporting System.<sup>26</sup> Nonviolent crime was limited to theft or criminal property crimes designated as "nonindex," which are defined by the National Incident-Based Reporting System as a "less serious offense."<sup>26</sup> Crime counts were aggregated to the census tract level and calculated as an annual crime rate per 1000 population; this rate was used as a proxy for individual exposure to crime. Census tracts were classified into 4 crime rate quartiles (low, medium, high, and very high) for each independent variable of interest (violent crime and nonviolent crime); each quartile contained 81 census tracts (Figure).

We additionally examined both the presence of and proximity to nearest homicide as independent variables, reasoning that homicide is an infrequent but potent violent crime (eg, high impact and media attention) and may be more



**Figure.** Distribution of crime types by census tract, South Side of Chicago, IL, from June 1, 2014 to May 31, 2015. \*Undefined signifies census tracts outside the CommunityRx study area.

likely to trigger psychological and physiological responses to violence (eg, stress and fear). Presence of homicide was defined as a dichotomous variable: no homicide or at least 1 homicide per census tract. Proximity to homicide was measured using Euclidean distance from each patient's residential address to the nearest homicide. Even in census tracts with no homicide, proximity to each homicide in the data set ( $n=269$ ) was considered, and distance to the nearest homicide was measured as a continuous value. Distances were subsequently categorized into 4 groups:  $>1$ , 0.5 to 1, 0.25 to 0.5, or  $<0.25$  miles from the nearest homicide.

Covariates included patient demographic characteristics, including age (continuous), sex (male or female), self-reported race/ethnicity (white non-Hispanic, black non-Hispanic, Hispanic or Latino, other race/ethnicity, and refused), and primary insurance status (private, Medicaid/Medicare, or none); clinic site (internal medicine, geriatrics, or gynecology); and neighborhood characteristics (2014 American Community Survey 5-year estimates),<sup>24</sup> including median household income (continuous),  $\geq 25\%$  less than high school graduation (dichotomous),  $\geq 25\%$  unemployed (dichotomous),  $\geq 75\%$  minority race/ethnicity (dichotomous), and  $\geq 10\%$  poor English fluency (dichotomous).

## Statistical Analysis

Descriptive statistics were used to characterize the demographic and census tract characteristics of our sample. We fit generalized linear mixed models (logit link function) to assess obesity and BP status as a function of crime rate quartile for each crime category (violent crime and nonviolent theft or property crime). All analyses accounted for clustering at the census tract level with the inclusion of a random effect for census tract.

We included all theoretically relevant confounders in final adjusted models, including aforementioned sociodemographic characteristics, clinic site, and neighborhood characteristics. Notably, we adjusted for neighborhood unemployment, which has been shown in prior literature to be independently linked to both high rates of crime and poor health outcomes.<sup>27,28</sup> Similarly, we adjusted for English fluency as a proxy for immigration status, paying special attention to the Mexican immigrant communities on Chicago's South Side. Recent attention on immigration in the United States has led to growing awareness that immigrants are more likely to be victims of violent crime, related to underreporting because of fear of law or immigration enforcement.<sup>29</sup> These same communities are often less likely to be insured,<sup>30</sup> less likely to access health care,<sup>31</sup> and more likely to have undiagnosed health conditions.<sup>32</sup> We tested for multicollinearity between neighborhood-level covariates using variance inflation factor; mean variance inflation factor was 1.50, with no values

exceeding the standard threshold of 10.<sup>33</sup> We additionally applied spline representations in a sensitivity analysis to adjust for age and median household income more flexibly; this analysis revealed no significant differences in the observed effect of violent crime for either obesity or elevated BP.

To further characterize the relationship between census tract-level exposure to crime and obesity or BP status, we conducted exploratory interaction tests and subgroup analyses by sex (male or female) and age group ( $<65$  or  $\geq 65$  years). The interaction effects were formally tested in the full model, with the reported sex-specific findings based on subgroup-restricted models. Statistical significance was defined as  $P<0.05$ .

We fit generalized linear mixed models to assess obesity and BP status as a function of census tract homicide (no homicide or at least 1 homicide) and proximity to homicide ( $>1$ , 0.5–1, 0.25–0.5, and  $<0.25$  miles). Model implementation was identical to that previously described, controlling for patient demographic characteristics, clinic site, and neighborhood characteristics.

Using empirical Bayes means estimation, we generated histogram and Q-Q plots to verify the assumption that the random effects were normally distributed. We also used bootstrap methods (500 replications) to examine robustness of the crime effect. Nonparametric bootstrap estimation was performed by resampling observations with replacement from the sample data in both a simple model and the fully specified model (mixed-effects model with adjustment for all confounders). These analyses validated our primary findings. Data were analyzed using Stata/SE, version 15.0 (StataCorp LP, College Station, TX) and GeoDa, version 1.10.0.12.

## Results

In a 324 census tract region of Chicago, median total crime rates in each quartile ranged from 54 to 240 crimes per 1000 population (Table 1). The quartile containing the *lowest* crime exposure had a median violent crime rate of 15 per 1000 population, similar to the national average rate of 19 per 1000 population.<sup>34</sup> The quartile containing the *highest* crime exposure had a median violent crime rate of 84 per 1000 population (Figure),  $>4$  times higher than the national average. Almost two thirds of census tracts (65%) had  $\geq 25\%$  residents living below the federal poverty level; more than a third (37%) had an unemployment rate of  $\geq 25\%$ ; and more than a third (38%) had a high school graduation rate of  $<75\%$ . More than half of census tracts (58.8%) were composed of a black non-Hispanic majority population ( $\geq 75\%$  of the population) (Table 1).

Of the 14 799 patients included in this study, 42% were obese and 33% had elevated BP (Table 2). Most patients were aged  $\geq 45$  years (67.5%), women (76.8%), and black non-

**Table 1.** Characteristics of Census Tracts, South Side of Chicago, IL

Demographic Composition(N=324 Census Tracts)*	Census Tracts, %
Below federal poverty level	
<25%	34.7
25%–49%	56.7
≥50%	8.7
High school graduation or higher	
<50%	5.0
50%–74%	32.5
75%–100%	61.6
Unemployment rate	
<25%	62.8
25%–49%	36.2
≥50%	1.0
Uninsurance rate	
<25%	74.3
25%–49%	25.7
≥50%	0
Black non-Hispanic	
<25%	28.2
25%–74%	13.0
75%–100%	58.8
Hispanic or Latino	
<25%	69.3
25%–74%	15.4
75%–100%	15.2
Census Tract Quartiles	Median Census Tract Crime per 1000 Population (IQR)
Total crime	
Low	54.0 (43.3–62.2)
Medium	109.9 (93.6–121.9)
High	162.1 (152.5–176.7)
Very high	240.3 (216.9–271.0)
Nonviolent theft or property crime	
Low	19.2 (15.7–21.7)
Medium	37.8 (33.1–40.6)
High	48.3 (45.2–52.1)
Very high	64.4 (58.3–76.5)
Violent crime	
Low	14.5 (11.7–17.1)
Medium	32.7 (26.5–37.8)
High	56.4 (48.8–60.0)
Very high	84.1 (78.5–99.3)

IQR indicates interquartile range.

\*Census tract data were abstracted from the 2014 American Community Survey (5-year estimates) and reflect a densely populated study region of 992,000 residents.

Hispanic (72.4%); half had Medicaid and/or Medicare (Table 2).

Compared with patients living in the lowest quartile, patients living in the highest quartile for violent crime rate had 53% higher adjusted odds of obesity (95% confidence interval [CI], 1.15–2.03) and 25% higher adjusted odds of elevated BP (95% CI, 1.01–1.56; Table 3). Patients living in the highest quartile for nonviolent theft or property crime had 41% higher adjusted odds of obesity (95% CI, 1.13–1.76); higher exposure to nonviolent theft or property crime was not significantly associated with BP status in adjusted models (Table 3).

**Table 2.** Patient Characteristics From Primary Care Clinics

Patient Characteristics (N=14 799)*	% of Patients
Age, y	
18–24	6.9
25–34	12.9
35–44	12.8
45–54	13.7
55–64	16.0
65–74	16.3
≥75	21.5
Sex	
Female	76.8
Male	23.2
Race	
White non-Hispanic	17.6
Black non-Hispanic	72.4
Hispanic or Latino	4.1
Other	5.0
Refused	0.9
Insurance status	
Private	44.1
Medicaid and/or Medicare	51.0
None or unknown	5.0
Weight status (BMI)	
Underweight (<18.5 kg/m <sup>2</sup> )	2.1
Normal weight (18.5–24.9 kg/m <sup>2</sup> )	27.6
Overweight (25–29.9 kg/m <sup>2</sup> )	28.4
Obese (≥30 kg/m <sup>2</sup> )	41.9
BP status, mm Hg	
Systolic BP <140 and diastolic BP <90	66.8
Systolic BP ≥140 or diastolic BP ≥90	33.2

BMI indicates body mass index; and BP, blood pressure.

\*Patient data were abstracted from clinical visits recorded in the electronic health record between June 1, 2014 and May 31, 2015.

**Table 3.** Census Tract Crime and BMI or BP Status

Crime Type	Obesity (BMI $\geq 30 \text{ kg/m}^2$ ) (n=12 358)*			Elevated BP (SBP $\geq 140$ or DBP $\geq 90 \text{ mm Hg}$ ) (n=14 331)†		
	%	OR (95% CI)	AOR (95% CI)‡	%	OR (95% CI)	AOR (95% CI)‡
<b>Nonviolent crime rate quartile</b>						
Low	24.8	Reference	Reference	24.1	Reference	Reference
Medium	44.2	1.89 (1.49–2.40)§	1.57 (1.27–1.93)§	32.8	1.33 (1.10–1.61)¶	1.09 (0.94–1.27)
High	44.9	1.79 (1.42–2.25)§	1.36 (1.10–1.68)¶	36.1	1.44 (1.20–1.73)§	1.14 (0.97–1.33)
Very high	48.7	1.82 (1.44–2.28)§	1.41 (1.13–1.76)¶	36.6	1.44 (1.20–1.73)§	1.15 (0.98–1.36)
<b>Violent crime rate quartile</b>						
Low	24.8	Reference	Reference	24.1	Reference	Reference
Medium	41.2	1.69 (1.35–2.10)§	1.36 (1.07–1.73)¶	33.1	1.41 (1.19–1.69)§	1.14 (0.95–1.37)
High	50.0	2.07 (1.66–2.58)§	1.52 (1.17–1.96)¶	38.6	1.71 (1.44–2.04)§	1.24 (1.02–1.51)¶
Very high	50.3	2.10 (1.68–2.62)§	1.53 (1.15–2.03)¶	36.2	1.51 (1.26–1.80)§	1.25 (1.01–1.56)¶
<b>Presence of homicide<sup>#</sup></b>						
No homicide	36.7	Reference	Reference	30.0	Reference	Reference
At least 1 homicide	49.1	1.37 (1.17–1.60)§	1.07 (0.95–1.21)	37.7	1.28 (1.14–1.45)§	1.16 (1.06–1.27)¶
<b>Proximity to nearest homicide, mile</b>						
>1	33.8	Reference	Reference	28.9	Reference	Reference
>0.5–1	34.4	1.32 (1.01–1.71)¶	0.93 (0.72–1.20)	29.4	1.16 (0.92–1.48)	0.98 (0.78–1.24)
>0.25–0.5	44.7	1.46 (1.11–1.91)¶	0.96 (0.73–1.25)	34.8	1.20 (0.94–1.53)	1.01 (0.80–1.29)
$\leq 0.25$	48.4	1.58 (1.19–2.08)¶	1.00 (0.76–1.32)	36.3	1.24 (0.96–1.59)	1.08 (0.84–1.38)

AOR indicates adjusted OR; BMI, body mass index; BP, blood pressure; CI, confidence interval; DBP, diastolic BP; OR, odds ratio; and SBP, systolic BP.

\*Derived from height and weight measurements recorded in the electronic health record; standard definitions obtained from the Centers for Disease Control and Prevention.

†Derived from systolic and diastolic BP measurements recorded in the electronic health record; standard definitions obtained from the Seventh Report of the Joint National Committee.

‡Estimates obtained from generalized linear mixed models. Adjusted for patient demographics, including age (continuous), sex, race, and insurance status; clinic site; and census tract characteristics, including median household income (continuous),  $\geq 25\%$  less than high school graduation,  $\geq 25\%$  unemployment,  $\geq 75\%$  minority racial/ethnic composition, and  $\geq 10\%$  poor English fluency.

§ $P<0.001$ .

¶ $P<0.01$ .

\* $P<0.05$ .

<sup>#</sup>Presence of homicide on the basis of census tract of residence.

In adjusted models examining continuous values for BMI and systolic BP, patients living in the highest quartile for violent crime had, on average, 1.6  $\text{kg/m}^2$  higher BMI (95% CI, 0.59–2.57  $\text{kg/m}^2$ ;  $P=0.002$ ) and 2.2 mm Hg higher systolic BP (95% CI, 0.32–4.02 mm Hg;  $P=0.02$ ) compared with those living in the lowest quartile (Table S1).

In interaction tests by sex, living in a neighborhood with higher rates of violent crime had a larger effect on obesity rates among women than among their male counterparts ( $\chi^2[3 \text{ df}] = 31.2$ ;  $P<0.001$ ); interaction tests were not statistically significant for elevated BP ( $\chi^2[3 \text{ df}] = 7.4$ ;  $P=0.06$ ). In subgroup analyses, women living in the highest quartile for violent crime had 71% higher adjusted odds of obesity (95% CI, 1.25–2.35) than those living in the lowest quartile (Table 4). In interaction tests by age group, the associations between violent crime and obesity ( $\chi^2[3 \text{ df}] = 28.0$ ;  $P<0.001$ ) or elevated BP ( $\chi^2[3 \text{ df}] = 16.2$ ;  $P=0.001$ ) were weaker among elderly adults (aged  $\geq 65$  years) than among their nonelderly

counterparts; however, effect sizes in adjusted models were similar between age groups in subgroup analyses (Table S2).

Homicides were rare events, ranging from 0 to 6 homicides per census tract. There were 13 “double homicides” (single events with 2 homicides), but there were no events with  $\geq 3$  homicides recorded during the study period. We were unable to determine if homicides occurring on the same date in different locations were related events. Similarly, there were homicides occurring in identical locations over a short period of time (eg, 5–10 days); we were unable to determine if these were related events (eg, retaliation). Most census tracts (52%) had no homicides during the study period; few census tracts (8%) had  $\geq 3$  homicides. Approximately a quarter of patients (25.7%) lived one-quarter mile or less from the nearest homicide. Compared with patients living in a census tract with no homicides, patients living in a census tract with at least 1 homicide had 16% higher adjusted odds of elevated BP (95% CI, 1.06–1.27; Table 3); in analyses examining obesity status,

**Table 4.** Subgroup Analysis by Sex: Census Tract Violent Crime and BMI or BP Status

Violent Crime by Sex Crime Rate Quartile	Obesity (BMI $\geq 30 \text{ kg/m}^2$ ) (n=12 358)*			Elevated BP (SBP $\geq 140$ or DBP $\geq 90 \text{ mm Hg}$ ) (n=14 331)†		
	%	OR (95% CI)	AOR (95% CI)‡	%	OR (95% CI)	AOR (95% CI)‡
<b>Men and women</b>						
Low	24.8	Reference	Reference	24.1	Reference	Reference
Medium	41.2	1.69 (1.35–2.10)§	1.36 (1.07–1.73)	33.1	1.41 (1.19–1.69)§	1.14 (0.95–1.37)
High	50.0	2.07 (1.66–2.58)§	1.52 (1.17–1.96)	38.6	1.71 (1.44–2.04)§	1.24 (1.02–1.51)
Very high	50.3	2.10 (1.68–2.62)§	1.53 (1.15–2.03)	36.2	1.51 (1.26–1.80)§	1.25 (1.01–1.56)
<b>Women only</b>						
Low	24.8	Reference	Reference	21.5	Reference	Reference
Medium	43.4	1.92 (1.49–2.46)§	1.55 (1.19–2.02)	31.6	1.55 (1.26–1.90)§	1.18 (0.95–1.46)
High	52.7	2.45 (1.91–3.14)§	1.75 (1.32–2.34)§	37.5	1.90 (1.55–2.33)§	1.28 (1.01–1.62)
Very high	52.6	2.37 (1.85–3.04)§	1.71 (1.25–2.35)	34.0	1.60 (1.30–1.97)§	1.23 (0.95–1.59)
<b>Men only</b>						
Low	24.8	Reference	Reference	30.8	Reference	Reference
Medium	32.7	1.39 (1.05–1.85)	1.01 (0.71–1.45)	38.2	1.35 (1.06–1.72)	1.07 (0.79–1.44)
High	39.5	1.76 (1.31–2.36)§	1.19 (0.80–1.77)	42.5	1.59 (1.25–2.02)§	1.13 (0.81–1.58)
Very high	40.4	1.85 (1.36–2.50)§	1.27 (0.82–1.95)	45.0	1.75 (1.36–2.27)§	1.32 (0.91–1.90)

AOR indicates adjusted OR; BMI, body mass index; BP, blood pressure; CI, confidence interval; DBP, diastolic BP; OR, odds ratio; and SBP, systolic BP.

\*Derived from height and weight measurements recorded in the electronic health record; standard definitions obtained from the Centers for Disease Control and Prevention.

†Derived from systolic and diastolic BP measurements recorded in the electronic health record; standard definitions obtained from the Seventh Report of the Joint National Committee.

‡Estimates obtained from generalized linear mixed models. Adjusted for patient demographics, including age (continuous), sex, race, and insurance status; clinic site; and census tract characteristics, including median household income (continuous),  $\geq 25\%$  less than high school graduation,  $\geq 25\%$  unemployment,  $\geq 75\%$  minority racial/ethnic composition, and  $\geq 10\%$  poor English fluency.

§ $P<0.001$ .

|| $P<0.05$ .

¶ $P<0.01$ .

findings were not statistically significant in adjusted models (adjusted odds ratio=1.07; 95% CI, 0.95–1.21). Proximity to nearest homicide was not significantly associated with obesity or BP status in adjusted models (Table 3).

## Discussion

We found that exposure to a higher police-recorded violent crime rate was consistently associated with measured obesity and elevated BP in this sample of urban-dwelling adults. This study, using objective measures of both violence and health (BMI and BP), corroborates prior work suggesting a relationship between neighborhood safety and health.<sup>35–37</sup> Higher exposure to nonviolent theft or property crime was also associated with obesity, but these effects were somewhat attenuated compared with analyses examining exposure to violent crime. In the most similar study published to date, Miranda and colleagues examined 1785 school-aged children and found a stronger relationship between violent crime rates and measured BMI, compared with property crime rates.<sup>38</sup> However, similar studies using objective measures in adult

populations have typically examined homicide rates only, resulting in null findings.<sup>11,39,40</sup> Our study distinguishes between 3 types of crime (violent crime, nonviolent property crime, and homicide), revealing notable differences in their associations with obesity and elevated BP. Findings in our study were also more pronounced among women, supporting prior evidence that women who feel unsafe may be less likely to engage in physical activity relative to men.<sup>7</sup> This finding, along with the high prevalence of female-headed households in neighborhoods with the most crime,<sup>41</sup> aligns with growing concern about disparities in cardiovascular outcomes among women residing in disadvantaged communities.<sup>22</sup>

Consistent with prior literature,<sup>11,39,40</sup> we additionally found that exposure to *homicide alone* was not associated with obesity status and was inconsistently associated with BP status. The presence of at least 1 homicide in a community was associated with elevated BP; however, the relative effect was smaller compared with analyses examining overall violent crime rates. Moreover, proximity to nearest homicide was not associated with obesity or BP status in adjusted models. Our findings possibly suggest that recurrent exposure to violent

crime may be more strongly associated with physical health than a single exposure to severe violent crime (ie, homicide). Interestingly, prior research has identified that most adults confronted with a serious life event (eg, death of a loved one) do not experience long-term deleterious health effects.<sup>42–44</sup> However, other research has revealed that exposure to an extreme event of violence (eg, a terrorist attack) can trigger poor physical health.<sup>45,46</sup> Our study offers preliminary evidence to support the hypothesis that secondary exposure to recurrent violence may confer unique health risks compared with isolated events of lethal violence. However, it is also possible that homicide has rippling effects that are more diffuse, both spatially and socially, which may not be captured in studies of neighborhood effects. For instance, robbery concerns may be neighborhood specific, whereas homicide may affect more diffuse social networks and settings.

Nevertheless, related research may point to some possible explanations for the observed differences between these types of violent crime. Studies of racial discrimination find that “everyday discrimination” is more strongly associated with physical health than rare events of severe discrimination, even for the most toxic experiences.<sup>47,48</sup> Prior studies have attributed these differences to the biological impacts of chronic stressors, or allostatic load, on autonomic regulation, immune and endocrine system function, and inflammatory processes.<sup>15,44</sup> Despite overwhelming public attention on homicides, it is possible that incremental and cumulative stress because of “everyday” crime (eg, robbery) plays a larger role in physical health. Another possibility is that persistent occurrences of nonlethal violence trigger behavioral responses that affect daily activities, such as reduced physical activity, whereas isolated events of homicide do not. Indeed, homicides may be perceived as being personal and targeted events, because only a minority of homicides are committed by strangers,<sup>49</sup> whereas robberies may be perceived as more impersonal and “random” events that can occur to anyone who appears vulnerable (eg, frail, obese, and disabled people). It is also possible that a single homicide may have few practical impacts on neighborhood businesses, whereas persistent robbery concerns may prompt businesses to avoid locating to high-crime neighborhoods.

Overall, our study of a predominantly black primary care population points to a relationship between exposure to violent crime and chronic disease status, with possible implications for clinical population health management. Although prior literature has described the importance of addressing community violence as a public health issue, healthcare systems may have financial incentive to address community violence as a *population* health issue. Indeed, policies that favor per-capita and performance-based payments may inadvertently tie reimbursement to population-level social conditions that determine health. Thus, healthcare

organizations located in regions of particular social vulnerability, similar to the region examined in this study, must think critically about broader approaches to optimize health. For instance, healthcare organizations located in high-crime regions may opt to invest in violence prevention and intervention strategies.<sup>50</sup> We speculate that these strategies may have potential health benefits for not only those directly affected by violence (eg, trauma victims), but also for those with downstream health consequences (eg, obesity, hypertension, and mental health). As such, recognizing and addressing the impacts of violence may be an important consideration for managing the health risks of low-income urban communities of color.

This study has several limitations. First, this is a cross-sectional study, which limits causal inference. Second, this study uses data from 3 primary care clinics at an academic medical center, limiting generalizability. For instance, half of the patients in our sample were insured by Medicaid and/or Medicare; our findings should be interpreted in light of these characteristics. For related reasons, we did not use diagnostic codes or prescription medications recorded in the electronic health record for classification purposes. Estimates for the completeness of diagnostic codes have been highly variable in prior studies,<sup>51</sup> and although prescription medications may have been used to delineate treated and untreated individuals, medication underuse has been reported to be high in this population.<sup>52</sup> We were concerned that these additional electronic health record data may have conferred imprecise subclassifications. By classifying BP on the basis of clinical measurement data alone, individuals without hypertension and individuals with controlled hypertension were analyzed as a single group. However, prior literature has documented that medication adherence is itself a function of safety.<sup>53</sup> Thus, classifications are consistent with hypotheses about both the physiological and behavioral health benefits of living in a low-crime neighborhood.

In addition, 16.5% of patients had missing BMI information; fewer (3.2%) had missing BP information. It is possible that those with missing BMI information declined or were unable to use the weight scale (eg, because of disability, wheelchair confinement, or related issues). In sensitivity analyses, however, patients with missing BMI information did not differ appreciably from those without missing BMI information (Table S3); also, sensitivity testing using multiple imputation methods did not alter the study findings (Table S4). Although we did not have data on individual socioeconomic characteristics, we partially mitigated this concern by analyzing a sample of individuals with access to health care, adjusting for person-level insurance type, and using multilevel models to adjust for neighborhood-level socioeconomic status indicators, similar to prior studies.<sup>37</sup> However, point estimates may be overestimated. In addition, the CommunityRx study

protocol abstracted health records during the study period only, and did not abstract health records before this period. Thus, longitudinal associations were not explored, and future studies should consider temporal trends.

We assigned exposure to crime using geographic methods, thus subjecting our findings to ecological fallacy: patients' level of exposure to violence, even within the same census tract, may vary depending on individual factors. Although census tract remains the enduring choice for neighborhood effects research because of relatively homogeneous population characteristics, economic status, and living conditions,<sup>54</sup> it is possible that patients were also affected by surrounding or more distant contexts, such as work environments or activity spaces. Furthermore, the objective secondary data used in this analysis do not enable assessments of individual trauma from violent events; therefore, we are limited in our ability to interpret findings within the broader literature on the psychological and behavioral effects of trauma. Finally, it is possible that the completeness of police crime records may vary geographically and by the severity of the crime, because of recording practices, reporting practices, or both. However, there is no compelling evidence for systematic discrepancies between census tracts. Similarly, violent crimes, such as homicide and assault, are likely to be known and recorded by police and, therefore, are likely to reflect accurate estimates of exposure.

## Conclusion

In a densely populated, high-poverty region in Chicago, exposure to a higher rate of violent crime was consistently associated with obesity and elevated BP among adult primary care patients. Moreover, a high rate of recurrent exposure to all types of violent crime was more strongly associated with health than a rare exposure to severe violent crime (homicide). Interventions to mitigate the harmful health effects of everyday violent crime, particularly in high-crime communities, may be a critical consideration for sustainably improving the cardiovascular health of vulnerable populations.

## Author Contributions

Study concept and design: Tung, Makelarski, Peek, and Lindau. Acquisition of data: Tung, Makelarski, and Lindau. Analysis and interpretation of data: Tung, Wroblewski, and Boyd. Drafting of the article: Tung, Makelarski, Peek, and Lindau. Critical revision of the article for important intellectual content: All authors. Statistical Analysis: Tung, Wroblewski, and Boyd. Administrative, technical, or material support: Peek and Lindau. Final approval of the version to be published: All authors. Tung had full access to all of the data in the study

and takes responsibility for the integrity of the data and the accuracy of the data analysis.

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## Disclosures

None.

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# **SUPPLEMENTAL MATERIAL**

**Table S1. Sensitivity Analysis of Continuous Outcomes: Census Tract Violent Crime and Body Mass Index or Systolic Blood Pressure.**

<b>Violent Crime</b> Crime Rate Quartile	<b>Body Mass Index (kg/m<sup>2</sup>)<sup>a</sup></b> n = 12,343		<b>Systolic Blood Pressure (mmHg)<sup>b</sup></b> n = 14,331	
	$\beta$ (95% CI)	Adjusted $\beta$ (95% CI) <sup>c</sup>	$\beta$ (95% CI)	Adjusted $\beta$ (95% CI) <sup>c</sup>
Low	Ref	Ref	Ref	Ref
Medium	1.72 (0.92, 2.52)***	1.14 (0.31, 1.97)**	3.31 (1.58, 5.05)***	1.02 (-0.53, 2.57)
High	2.61 (1.82, 3.40)***	1.60 (0.69, 2.50)**	5.36 (3.64, 7.08)***	1.92 (0.22, 3.61)*
Very High	2.62 (1.83, 3.42)***	1.58 (0.59, 2.57)**	4.16 (2.42, 5.90)***	2.17 (0.32, 4.02)*

\*P < 0.05, \*\*P < 0.01, \*\*\*P < 0.001

<sup>a</sup>Derived from height and weight measurements recorded in the Electronic Health Record (EHR); standard definitions obtained from the Centers for Disease Control and Prevention

<sup>b</sup>Derived from systolic and diastolic blood pressure measurements recorded in the EHR; standard definitions obtained from the Seventh Report of the Joint National Committee

<sup>c</sup>Estimates obtained from generalized linear mixed models; adjusted for patient demographics, including age (continuous), sex, race, and insurance status; clinic site; and census tract characteristics, including median household income (continuous), ≥25% less than high school graduation, ≥25% unemployment, ≥75% minority racial/ethnic composition, and ≥10% poor English fluency

**Table S2. Subgroup Analysis by Age Group: Census Tract Violent Crime and Body Mass Index or Blood Pressure Status.**

<b>Violent Crime by Age Group</b>	<b>Obesity</b> (BMI $\geq 30$ ) <sup>a</sup> n = 12,358			<b>Elevated Blood Pressure</b> (SBP $\geq 140$ or DBP $\geq 90$ ) <sup>b</sup> n = 14,331		
	Crime Rate Quartile	%	OR (95% CI)	AOR (95% CI) <sup>§</sup>	%	OR (95% CI)
All Ages						
Low	24.8	Ref	Ref	24.1	Ref	Ref
Medium	41.2	1.69 (1.35, 2.10)***	1.36 (1.07, 1.73)*	33.1	1.41 (1.19, 1.69)***	1.14 (0.95, 1.37)
High	50.0	2.07 (1.66, 2.58)***	1.52 (1.17, 1.96)**	38.6	1.71 (1.44, 2.04)***	1.24 (1.02, 1.51)*
Very High	50.3	2.10 (1.68, 2.62)***	1.53 (1.15, 2.03)**	36.2	1.51 (1.26, 1.80)***	1.25 (1.01, 1.56)*
Age less than 65 years						
Low	25.5	Ref	Ref	15.5	Ref	Ref
Medium	43.6	1.82 (1.37, 2.42)***	1.27 (0.95, 1.70)	22.6	1.58 (1.28, 1.95)***	1.18 (0.93, 1.50)
High	58.5	2.52 (1.90, 3.35)***	1.51 (1.09, 2.08)*	28.4	1.97 (1.59, 2.43)***	1.27 (0.98, 1.66)
Very High	57.0	2.51 (1.89, 3.33)***	1.53 (1.07, 2.17)*	26.9	1.81 (1.46, 2.23)***	1.28 (0.96, 1.71)
Age 65 years and older						
Low	23.3	Ref	Ref	43.0	Ref	Ref
Medium	37.2	1.84 (1.45, 2.34)***	1.51 (1.11, 2.06)**	50.8	1.29 (1.07, 1.57)**	1.11 (0.86, 1.43)
High	39.3	1.99 (1.57, 2.52)***	1.54 (1.11, 2.14)*	52.2	1.40 (1.17, 1.69)***	1.16 (0.88, 1.53)
Very High	39.1	1.96 (1.53, 2.51)***	1.60 (1.12, 2.29)*	52.2	1.38 (1.13, 1.68)**	1.21 (0.90, 1.63)

\*P < 0.05, \*\*P < 0.01, \*\*\*P < 0.001

<sup>a</sup>Derived from height and weight measurements recorded in the Electronic Health Record (EHR); standard definitions obtained from the Centers for Disease Control and Prevention

<sup>b</sup>Derived from systolic and diastolic blood pressure measurements recorded in the EHR; standard definitions obtained from the Seventh Report of the Joint National Committee

<sup>§</sup>Estimates obtained from generalized linear mixed models; adjusted for patient demographics, including age (continuous), sex, race, and insurance status; clinic site; and census tract characteristics, including median household income (continuous),  $\geq 25\%$  less than high school graduation,  $\geq 25\%$  unemployment,  $\geq 75\%$  minority racial/ethnic composition, and  $\geq 10\%$  poor English fluency

**Table S3. Patient Characteristics by Missing or Non-Missing BMI Information.**

Patient characteristics*	Non-Missing (n=12,369)		Missing (n=2,414)	
	Mean	%	Mean	%
Age (years)	55.6		55.7	
Female sex				
Clinic 1	76.6		78.3	
Clinic 2	100.0		100.0	
Clinic 3	66.0		66.6	
Black non-Hispanic	72.7		70.5	
Medicaid and/or Medicare	51.5		48.1	
Violent Crime Quartile				
Low	22.2		24.8	
Medium	29.1		29.1	
High	26.5		25.2	
Very high	22.2		20.9	

\*Patient data were abstracted from clinical visits recorded in the Electronic Health Record between June 2014 and May 2015.

**Table S4. Sensitivity Analysis: Imputed for Missing Body Mass Index.**

Crime Type Crime Rate Quartile	%	<u>Obesity</u> (BMI $\geq$ 30) <sup>†</sup> n = 14,783	
		OR (95% CI)	AOR (95% CI) <sup>‡</sup>
Violent Low	29.3	Ref	Ref
Medium	42.8	1.56 (1.28-1.89)***	1.25 (1.01-1.55)*
High	50.3	1.86 (1.53-2.24)***	1.37 (1.09-1.72)**
Very High	50.8	1.88 (1.55-2.28)***	1.38 (1.08-1.76)*

\*P < 0.05, \*\*P < 0.01, \*\*\*P < 0.001

<sup>†</sup>Derived from height and weight measurements recorded in the Electronic Health Record (EHR); standard definitions obtained from the Centers for Disease Control and Prevention

<sup>‡</sup>Estimates obtained from generalized linear mixed models; adjusted for patient demographics, including age (continuous), sex, race, and insurance status; clinic site; and census tract characteristics, including median household income (continuous),  $\geq$ 25% less than high school graduation,  $\geq$ 25% unemployment,  $\geq$ 75% minority racial/ethnic composition, and  $\geq$ 10% poor English fluency