Contents lists available at ScienceDirect

Drug and Alcohol Dependence

journal homepage: www.elsevier.com/locate/drugalcdep

County-level predictors of US drug overdose mortality: A systematic review

Manuel Cano ^{a,*}, Sehun Oh^b, Preston Osborn^b, Samson A. Olowolaju^c, Armando Sanchez^d, Yeonwoo Kim^{e, f}, Alberto Cano Moreno^g

^a School of Social Work, Arizona State University, USA

^b College of Social Work, The Ohio State University, USA

^c Department of Demography, University of Texas at San Antonio, USA

^d Department of Social Work, University of Texas at San Antonio, USA

^e Department of Kinesiology, University of Texas at Arlington, USA

^f School of Social Work, University of Texas at Arlington, USA

^g Department of Public Policy, Universidad Autónoma del Estado de Hidalgo, México

ARTICLE INFO

Keywords: Drug overdose County-level Systematic review Opioid Mortality

ABSTRACT

Background: This systematic review summarized published literature on county-level predictors of drug overdose mortality in the United States (US).

Methods: Peer-reviewed studies and doctoral dissertations published in English between 1990 and July 19, 2022 were identified from PubMed, Web of Science, ProQuest Dissertations & Theses, PsycINFO, CINAHL, and EconLit. Eligible studies examined at least one county-level predictor of drug overdose mortality in US counties. Two reviewers independently completed screening, quality assessment (with an adapted National Institutes of Health Quality Assessment Tool), and data extraction. Results were qualitatively summarized and grouped by predictor categories.

Results: Of 56 studies included, 42.9% were subnational, and 53.6% were limited to opioid overdose. In multiple studies, measures related to opioid prescribing, illness/disability, economic distress, mining employment, incarceration, family distress, and single-parent families were positively associated with drug overdose mortality outcomes, while measures related to cannabis dispensaries, substance use treatment, social capital, and family households were negatively associated with drug overdose mortality outcomes. Both positive and negative associations were documented for smoking, uninsurance, healthcare professional shortage status, physicians per capita, unemployment, income, poverty, educational attainment, racial composition, and rurality. Findings within studies also differed by subpopulation (by race/ethnicity, gender, age, or rurality) and the type of drugs involved in overdose.

Conclusions: The findings of this review provide relatively mixed evidence regarding many county-level predictors of overdose mortality, several of which also vary between subpopulations, supporting the importance of additional research to elucidate pathways through which the county context may shape risk of fatal overdose.

1. Introduction

Drug overdose deaths have increased four-fold in the United States (US) since 1999 (Centers for Disease Control and Prevention, 2021), exacting a staggering economic and societal burden (Florence et al., 2021; Gomes et al., 2018). Although the escalation in overdose deaths has impacted all regions across the US, researchers have documented wide geographic variations in rates of drug overdose mortality (Rossen et al., 2014). For example, age-adjusted drug overdose mortality rates in

2019 ranged from 4.8 deaths per 100,000 residents in Hidalgo County, Texas to 120.1 deaths per 100,000 residents in Scioto County, Ohio (Centers for Disease Control and Prevention, 2022).

Numerous studies have examined potential underlying causes of this geographic variation in US drug overdose mortality (Langabeer et al., 2020; Monnat, 2018, 2019; Monnat et al., 2019), yet such studies vary in methodology, measures, and specific outcomes examined. The present systematic review aims to synthesize the knowledge base regarding county-level factors associated with drug overdose mortality across the

* Corresponding author. *E-mail address:* m.cano@asu.edu (M. Cano).

https://doi.org/10.1016/j.drugalcdep.2022.109714

Received 18 May 2022; Received in revised form 21 November 2022; Accepted 21 November 2022 Available online 24 November 2022 0376-8716/© 2022 Elsevier B.V. All rights reserved.



Review





US. The review's focus on counties, as opposed to broader geographic levels (e.g., states), is informed by the substantial extent of within-state variation in drug overdose mortality rates. The county represents the smallest geographic unit available in national mortality data (Centers for Disease Control and Prevention, 2022); various subnational studies have examined drug overdose mortality at more granular levels (e.g., neighborhoods, cities/towns, or zip code tabulation areas; Cerdá et al., 2013; Green et al., 2011; Hester et al., 2012), yet data at these levels are not routinely available for the entire US.

Literature reviews related to fatal or non-fatal drug overdoses have been published on a variety of topics, including (but not limited to) prevalence and trends (Martins et al., 2015), risk factors for overdose in general or specific subpopulations (Armoon et al., 2021a, 2021b; Brady et al., 2017; Flam-Ross et al., 2022; Larney et al., 2020; Lyons et al., 2019; Mital et al., 2020; Van Draanen et al., 2020, 2021), and interventions for overdose prevention (Bahji and Bajaj, 2018; Ma et al., 2019; McDonald and Strang, 2016; Mercer et al., 2021; Potier et al., 2014). Most of these reviews have focused on research about drug overdose at the individual level (Armoon et al., 2021a, 2021b; Brady et al., 2017; Flam-Ross et al., 2022; Larney et al., 2020; Ma et al., 2019; Mital et al., 2020; Van Draanen et al., 2021), while others have included research at the state level (Beaudoin et al., 2016) or multiple levels (Van Draanen et al., 2020). The present systematic review, in contrast, focuses exclusively on county-level predictors of drug-related mortality, as factors at levels beyond the individual represent opportunities for interventions with broader reach and impact (Frieden, 2010).

2. Methods

This systematic review was based on guidelines from the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA), as detailed subsequently. (The review was not registered, and no protocol was published.) Eligible studies modeled at least one county-level indicator as a predictor or correlate of county-level rates of drug overdose mortality. We included studies that examined drug overdose mortality overall, as well as those that focused on deaths attributed to a specific drug type (e.g., opioids). Although the focus of the review was drug overdose mortality, we also included studies that used the broader designation of "drug-induced" ("drug-related") mortality, as drug overdoses account for approximately 95% of the deaths in this category (Centers for Disease Control and Prevention, 2022). However, studies that combined drug overdose deaths with chronic substance-related causes of death, such as alcoholic liver disease, were not eligible for inclusion. We included studies examining accidental/unintentional drug overdoses, considering that the majority (88%) of drug overdose deaths in the US are classified as accidental/unintentional (Centers for Disease Control and Prevention, 2022). Additionally, due to uncertainty in the classification of overdose intent (Rockett et al., 2018), we also included studies that combined accidental/unintentional drug overdoses with intentional drug overdoses or drug overdoses of undetermined intent.

Studies were eligible for inclusion if the setting comprised counties in the US, whether the entire US or a specific area within the US, with a population of either the overall resident population or a specific subgroup (e.g., certain age range or racial/ethnic group). Peer-reviewed articles and doctoral dissertations were eligible for inclusion, while master theses, non-peer-reviewed conference papers, and gray literature were excluded. In cases in which both a doctoral dissertation and its corresponding peer-reviewed publications met criteria for inclusion, the peer-reviewed publications were included while the doctoral dissertation was excluded, in order to avoid duplication. Only articles published in English were included. Considering that the contemporary US "opioid crisis" began in the 1990s (Centers for Disease Control and Prevention, 2021), articles were included if published after 1990 (and prior to the final literature search date of July 19, 2022).

2.1. Search Strategy

Studies were retrieved on September 23, 2020 and July 19, 2022 from six electronic databases (PubMed, Web of Science, ProQuest Dissertations & Theses, PsycINFO, CINAHL, and EconLit). The search terms used for each database are provided in Supplemental Table 1 and summarized as: (county or counties) and (overdos* or death* or mortalit* or fatal* or poisoning or decedent* or died) and (drug* or opioid* or opiat* or heroin or fentanyl or cocaine or methamphetamine* or stimulant* or psychostimulant* or benzodiazepine*).

Citations retrieved from all databases were consolidated in EndNote[™] and uploaded in the Covidence software, where duplicate entries were automatically removed. Additionally, citations from the reference lists of studies which met inclusion criteria in the first search (retrieved via the online Citationchaser tool: Haddaway et al., 2022) were also uploaded in Covidence for screening. Two reviewers (M.C. and P.O., S.A.O., A.S., or A.C.M.) independently evaluated each title/abstract to exclude any publications unrelated to the review topic. Any discrepancies were resolved with the input of a third independent reviewer (Y.K. or S.O.). Finally, two reviewers (M.C. and S.O. or A.C.M.) independently completed a full text review to determine whether each study met inclusion criteria, with discrepancies again resolved by a third independent reviewer (Y.K.). For the title/abstract screening, the mean Cohen's Kappa across pairs of reviewers was 0.54, suggesting "moderate" agreement (Landis and Koch, 1977, p. 165). For the full text review, the mean Cohen's Kappa across two pairs of reviewers was 0.93, suggesting "almost perfect" agreement (Landis and Koch, 1977, p. 165).

2.2. Data extraction and quality assessment

Two authors (M.C. and S.O. or A.C.M.) independently completed a quality assessment and data extraction form for each study. Any discrepancies in responses were resolved via discussion. The data extraction form, available in Supplemental Table 2, included information on the study's population, sample size, outcome measures, predictors, covariates, years, data sources, missing data, analytical plan, and results.

Considering that the review included studies examining a wide variety of predictors and several different types of drug-related mortality outcomes, we focused on extracting data regarding the direction of statistically significant associations between county-level predictors and drug-related mortality, without reporting measures of the strength of these associations. For studies that examined both county-level and noncounty-level (such as state-level) predictors, we extracted findings regarding the county-level predictors only. We focused on reporting results regarding associations with variables modeled as predictors, rather than also reporting associations with variables modeled as controls/covariates, in consideration of the risk of the "Table 2 Fallacy" (Westreich and Greenland, 2013). We provided result summaries for each study in table format and additionally qualitatively synthesized findings in diagrams based on predictor categories (health-related, economic or social factors, and demographic or geographic characteristics).

The quality assessment form used in the review is provided in Supplemental Table 3 and was adapted from the National Institutes of Health (NIH) Quality Assessment Tool for Observational Cohort and Cross-Sectional Studies (National Institutes of Health, 2021). The NIH Quality Assessment Tool has been previously used in systematic reviews of opioid overdose (Van Draanen et al., 2020, 2021) and was modified for the current review in order to accommodate studies at the county level. For example, questions regarding sample recruitment, blinded exposure status, participation rate, and loss to follow-up were not applicable to county-level analyses and were therefore eliminated.

In addition to the questions included in the quality assessment form, we also extracted data regarding a unique data quality concern for studies examining drug-specific overdose mortality at the county level. Researchers from the Centers for Disease Control and Prevention have highlighted the need for caution when examining opioid overdose mortality rates at the county level, in light of dramatic differences between counties in the completeness of drug identification and reporting on death certificates (Jones et al., 2019). Therefore, we extracted data on whether each study included analyses with drug-specific outcomes (e.g., opioid overdose deaths), which are subject to underreporting. For studies examining drug-specific outcomes, we also extracted data on whether (and how) authors reported addressing underclassification of drug type, such as by using imputation procedures (Boslett et al., 2020; Ruhm, 2018) or adding analyses for drug overdoses as an overall category.

3. Results

3.1. Identified studies

As depicted in Fig. 1, title/abstract screening was completed for 5659 studies; 5501 records were excluded after title/abstract review, primarily due to research areas unrelated to the systematic review topic (e. g., motor vehicle fatalities, cancer mortality, drug-resistant tuberculosis). Of the remaining 158 studies included in the full-text screening,

102 were excluded, rendering a final sample of 56 studies. Examples of studies that did not meet the inclusion criteria are available in Supplemental Table 4. Of the 56 studies included in the review, all were published in peer-reviewed journals, with the exception of three doctoral dissertations (Hall, 2022; Sun, 2021; Suriaga, 2021). All studies were published between 2013 and 2022, with 91% published between 2018 and 2022.

3.2. Study characteristics

Table 1 provides information regarding the population, setting, sample size, and excluded/ missing data for each study included in the review, while Table 2 details the outcome measures, predictors, covariates, and statistical methods reported. As presented in Table 1, seventeen studies in the review (Blake-Gonzalez et al., 2021; Cordes, 2018; Gunn et al., 2018; Haley et al., 2019; Hall, 2022; Henry et al., 2021; Jones et al., 2020; Kline et al., 2019; Lister et al., 2020; Marotta et al., 2019; Romeiser et al., 2019; Rowe et al., 2022; Sauber-Schatz et al., 2013; Sawyer et al., 2021; Suriaga, 2021; Valentini and Jaya-wardhana, 2019; Whitley et al., 2022) were limited to one particular state, while seven other studies were limited to a specific region such as



Fig. 1. PRISMA diagram for study identification, screening, and inclusion.

Pierce and Schott

(2020)

resident population overall

and by age category, race,

and sex

Table 1

Information on population/subpopulations, counties, and missing data for the 56 articles included in the systematic review.

| | | , 0 | 5 | |
|----------------------------------|---|--|--|--|
| Author (year) | Population/Subpopulation (s) Included | Area Included | N Counties or County-Years | Excluded or Missing Data |
| Betz and Jones | resident population overall | US; overall and metro/nonmetro | 26,337 nonmetro and 13,649 | n.s. |
| Blake-Gonzalez | resident population | VA | 84 county-equivalents | counties w/ missing outcome data excluded |
| Bradford and Bradford (2020) | resident population | US except AK, AR, ND, SD; overall | 32,269 county-years of | AK, AR, ND, SD excluded due to missing data on predictor |
| Bruch et al. (2021) | resident population | US-Mexico border | 21 counties; 43 in secondary | n.s. |
| Congdon (2020) | resident population | US; overall and by geographic | 3141 counties | overdose rates for counties w/ less than 10 deaths missing, predicted values modeled |
| Condex (2018) | underst a new lation | NC | - | acatis missing, predicted values modeled |
| Dean and Kimmel | resident population | US | 820 counties | n.s. excluded counties with suppressed mortality |
| Frankenfeld and Leslie (2019) | resident population overall and by race | US | 2067 counties overall; 3142 in imputed dataset: smaller N for | multiple imputation used for the 34% of counties with 10 or fewer deaths |
| Cabriel et al | Non Hisponia White | AL AD EL CA LA ME NO SO | race-specific analysis | evaluad acustica missing acustication of outcome |
| (2021) | population | TN | 738 counties | data |
| Griffith et al. (2021) | resident population | US | 26,970 county-years | two counties excluded; multiple imputation for missing values on the outcome; linear interpolation for missing values on covariates |
| Grigoras et al. (2018) | resident population | US | 832 counties for opioid mortality analysis; lower N for specific opioids | excluded counties w/ suppressed mortality data |
| Gunn et al. (2018) | resident population | NC | 100 counties | n.s. |
| Hadland et al. (2019) | resident population | US | 9398 county-years | multiple imputation for 6–13% of counties missing prescribing rates |
| Haley et al. (2019) | resident population | NY | 62 counties | n.s. |
| Hall (2022) | resident population | ТХ | 91 counties | counties without mortality data available excluded |
| Henry et al. (2021) | resident population | CA | 58 counties | n.s. |
| Hollingsworth et al. | resident population overall | US | 3138 counties; lower for race | n.s |
| (2017) | and by race | | analyses | |
| Hsu and Kovacs (2021) | resident population ages 21+ | 23 states w/ legal cannabis dispensaries, plus DC | 812 counties in main analyses | n.s. |
| Jones et al. (2020) | resident population | NC | n.s. | excluded unspecified number of counties w/ unavailable data |
| Kelly et al. (2021) | children under 12 years old | US | 3140 counties | excluded two smallest counties |
| Kerry et al. (2016) | resident population | US overall, West, and UT | n.s. | Poisson kriging used for counties with missing data on the outcome |
| Kerry et al. (2019) | resident population | Western US overall and "high risk" western states | n.s. | Poisson kriging used for counties with missing data on the outcome |
| Kline et al. (2019) | resident population | OH | 88 counties | unspecified for mortality analysis |
| Kurani et al. (2020) | resident population | US | 3133 counties | single imputation for unspecified number of counties missing predictor data |
| Langabeer et al. (2020) | resident population | US | 1058 counties | excluded counties with less than 10 deaths |
| Lee et al. (2019) | resident population | US | n.s. | n.s. |
| Lin et al. (2020) | resident population overall and by age category, sex, race, and educational attainment | US | 3132 counties | data on one covariate missing for 7 counties across 2 years |
| Lister et al. (2020) | resident population | MI | 83 counties | n.s. |
| Marotta et al. (2019) | resident population | NY | 62 counties | n.s. |
| Monnat (2018) | resident population | US | 3106 counties | multiple imputation for 623 counties (20%) with suppressed mortality data |
| Monnat (2019) | Non-Hispanic White | 48 contiguous states and DC; | 3047 counties | excluded Broomfield County and counties with |
| | population | overall and for urban and rural statuses | | fewer than 1000 NH White adults; multiple imputation used for 295 counties (9.7%) missing opioid prescribing data |
| Monnat et al. (2019) | resident population | US- 48 contiguous states and DC | 3079 counties | multiple imputation for 6–13% of counties missing prescribing data |
| Nosrati et al. (2019) | resident population | US | 2640 counties | AK, CT, DE, HI, RI, VT excluded due to no local jail system; 77 counties (including all in VA) excluded due to changing county boundaries over time; ~15% of observations excluded due to data quality concerns on predictor |
| Ombach et al. (2019) | resident population overall and by sex | US- contiguous states | 1116 counties; fewer for sex- specific analyses | counties with fewer than 20 prescription opioid overdose deaths excluded |

US

specific analyses 3122 counties

n.s.

Table 1 (continued)

| Author (year) | Population/Subpopulation (s) Included | Area Included | N Counties or County-Years | Excluded or Missing Data |
|---|---------------------------------------|--|----------------------------------|--|
| Rangachari et al. (2022) | resident population | US | 554 counties | counties without outcome data available excluded |
| Romeiser et al. (2019) | resident population | NY | n.s. | excluded one county w/ missing data on predictor |
| Rowe et al. (2022) | resident population | CA | 58 counties | n.s. |
| Rudolph et al | resident population overall | US except AK: only counties w/ | 2659 counties in labor force | for unemployment analysis excluded counties |
| (2020) | and by sex and race/ | unemployment above 4.9% or | analysis: 1701 counties in | w/ < -4.0% unemployed adults: for labor force |
| (2020) | and by sex and race/ | labor for a second stration | anarysis, 1701 counties in | $W/ \leq -4.5\%$ the inployed addits, for labor force |
| | ethnicity | labor-lorce nonparticipation | unemployment analysis; lewer for | analysis, excluded counties with $< =35\%$ adults |
| | | above 33% | subgroup-specific analyses | out of labor force; for subgroup analyses, excluded counties $w/ < 5000$ residents in |
| B 1 (0010) | | | | subgroup |
| Ruhm (2019) | resident population overall | US | 3098 counties | 29 counties excluded due to missing data on |
| | and by sex, race, education | | | predictors/covariates; education missing for |
| | level, and age group | | | \sim 5% of death certificates |
| Sauber-Schatz et al. (2013) | resident population | FL | 67 counties | n.s. |
| Sawyer et al. (2021) | resident population | IN | 92 counties | n.s. |
| Smith (2020) | adults only, overall and by | US- continental states except CA | 48.799 county-years | CA excluded due to dispensaries open prior to |
| | sex and race (NH White) | ••• ••••••••••••••••• | | study period |
| Sup (2021) | resident nonulation | US contiguous states, everall and | 2107 counties | lineer internelation or multiple imputation used |
| 5uii (2021) | resident population | be such as (much states, overall and | 5107 counties | Generation data an extreme and an distant |
| | | by urban/rural status | | for missing data on outcome and predictor/ |
| | | | | covariates |
| Suriaga (2021) | older adults (ages 65+) | FL, overall and by metro and nonmetropolitan status | 67 counties | 0-0.004% missing data reported |
| Swensen (2015) | resident population overall | US- contiguous states; overall, | 2408 counties | excluded counties without at least one |
| | and by age category, race. | urban and rural, and by income | | treatment facility |
| | and sex | tercile | | |
| Tacheva and Ivanov | resident nonulation | US | 2801 counties | missing outcome data addressed using interval |
| | resident population | 03 | 2091 counties | inissing outcome data addressed using interval |
| (2021) | | | | regression with missing (suppressed) values in |
| | | | | lowest interval |
| Valentini and Jayawardhana (2019) | resident population | GA | 159 counties | n.s. |
| Venkataramani | ages 18–65 only, overall and | counties in 30 commuting zones | 112 counties | n.s. |
| et al. (2020) | by age category, sex, and | with top quintile of | | |
| | race | manufacturing workers, mostly in | | |
| | Tucc | South / Midwest | | |
| TATI tel | | South/ Midwest | 10 | |
| (2022) | resident population | ОН | 19 counties | n.s. |
| Wu and Evangelist | ages 25–54, overall and by | US | 3137 counties | n.s. |
| (2022) | race, sex, and age | | | |
| Yang et al. (2021) | resident population | US | 2648 counties | excluded counties w/ missing values on predictors |
| Zemore et al | resident population | US-Mexico border states (CA_NM | 360 counties | counties with fewer than 10 deaths excluded |
| (2022) | restacht population | A7 TY) | coo countres | countro with lewer than to deaths excluded |
| (2022) | resident population | UE or or all and by minol (urban | 2024 counting | 100 counting avaluated due to missing data or |
| Ziiu et al. (2022) | resident population | os overall allu by rural/urban | 5054 counties | Too counties excluded due to missing data on |
| | | status | | predictors; Poisson-gamma model used for |
| | | | | estimating missing values on outcome variable |
| Zoorob and Salemi | resident population | US | 3104 counties | 31 counties (1%) excluded due to missing data |
| (2017) | | | | on the predictor |
| Zoorob (2018) | resident population | US | 9106 county-years | 12 county-years missing mortality data and one |
| | * * | | | county-year missing demographic data |

Abbreviations. AK, Alaska; AL, Alabama; AR, Arkansas; AZ, Arizona; CA, California; CT, Connecticut; DC, District of Columbia; DE, Delaware; FL, Florida; GA, Georgia; HI, Hawaii; IN, Indiana; LA, Louisiana; MI, Michigan; MS, Mississippi; N, number; NC, North Carolina; ND, North Dakota; NH, New Hampshire; NM, New Mexico; n.s., not specified; NY, New York; OH, Ohio; RI, Rhode Island; SD, South Dakota; TN, Tennessee; TX, Texas; US, United States; UT, Utah; VT, Vermont; VA, Virginia; w/, with.

the US-Mexico border (Bruch et al., 2021; Zemore et al., 2022), the western US (Kerry et al., 2019), southern states (Gabriel et al., 2021), or areas that met a specific characteristic (i.e., manufacturing share [Venkataramani et al., 2020]; unemployment or labor-force nonparticipation rates [Rudolph et al., 2020]; or legalized cannabis dispensaries [Hsu and Kovacs, 2021]); the other 32 studies focused on the US overall. Two studies (Gabriel et al., 2021; Monnat, 2019) examined overdose mortality in the Non-Hispanic White population only, and six studies (Hsu and Kovacs, 2021; Kelly et al., 2021; Smith, 2020; Suriaga, 2021; Venkataramani et al., 2020; Wu and Evangelist, 2022) were limited to specific age ranges; all other studies included the overall resident population.

As presented in Table 2, more than half of the studies (53.6%) were limited to overdose deaths involving opioids, in main analyses (Betz and

Jones, 2018; Cordes, 2018; Dean and Kimmel, 2019; Gabriel et al., 2021; Griffith et al., 2021; Grigoras et al., 2018; Gunn et al., 2018; Hadland et al., 2019; Haley et al., 2019; Hall, 2022; Henry et al., 2021; Hsu and Kovacs, 2021; Jones et al., 2020; Kline et al., 2019; Langabeer et al., 2020; Lee et al., 2019; Lister et al., 2020; Ombach et al., 2019; Rangachari et al., 2022; Romeiser et al., 2019; Rowe et al., 2022; Sawyer et al., 2021; Smith, 2020; Sun, 2021; Suriaga, 2021; Tacheva and Ivanov, 2021; Venkataramani et al., 2020; Wu and Evangelist, 2022; Yang et al., 2021; Zhu et al., 2022), while one study examined overdoses involving a variety of specific drug types, including stimulants (Bradford and Bradford, 2020), and the remaining studies included drug overdose deaths overall (involving any drug, with or without additional analyses for specific drugs). Variables modeled as predictors included measures related to *health* (e.g., opioid prescribing, other substance use,

Table 2

Information on measures and statistical analyses reported by the 56 articles included in the systematic review.

| Author (year) | Outcome Measure | Years for Outcome Measure | County-Level Predictors | Controls/Covariates | Analysis |
|---|---|---------------------------------|--|--|--|
| Betz and Jones (2018) Blake-Gonzalez et al. (2021) | opioid overdose deaths of any intent drug overdose deaths | 1999-2014 2008-2017 | employment and wage growth in different industries opioid prescribing; market share of 5 largest pharmacy suppliers; income; unemployment; mining employment; % with disability; % insured; population density; % high school or more; mean work commute: incarceration: % white | total employment; population; income; poverty none | regression using Bartik instrument semi-log period, fixed effects estimating equation |
| Bradford and Bradford (2020) | poisoning deaths involving: any opioid; prescription opioids; heroin; synthetic opioids; cocaine; stimulants with abuse potential; benzodiazepines; antidepressants; alcohol | 2004–2016 | eviction rate | income; % in poverty; unemployment; % male; % aged 20–64; % Caucasian; rurality; % uninsured; # of doctors per population; electronic PDMP (state level) | linear model with county fixed effects using a control function (2SRI) |
| Bruch et al. (2021) | drug overdose deaths | 1990–2017 | border wall construction | race/ethnicity, sex, and age composition; urbanization | semiparametric OLS difference-in-difference with county and year fixed effects |
| Congdon (2020) | drug overdose deaths of any intent | 2015–2017 | unemployment rate; unemployment rate change; Index of Concentration at Extremes (income inequality); social capital; urban-rural status; racial segregation; [opioid prescribing and fentanyl exposure measured at state level] | none | Bayesian Poisson-lognormal regression |
| Cordes (2018) | opioid involved overdose deaths, unintentional or intentional, involving: any opioid; heroin; methadone; synthetic opioids; semi- synthetic opioids; prescription opioids | 1999–2015 | median age; racial/ethnic composition; urban-rural status; poverty rate; unemployment rate; % with disability; uninsurance rate; educational attainment; per capita income | none | OLS regression |
| Dean and Kimmel (2019) | opioid-related overdose deaths of any intent, involving any type of opioid | 1999–2015 | trade-related job losses | population; income; racial demographics; population density; state-level labor union density; opioid prescription rates; number of physicians certified to prescribe buprenorphine; unemployment, mass layoffs unrelated to trade; presence of fentanyl | OLS regression |
| Frankenfeld and Leslie (2019) | drug overdose deaths of any intent | 2010–2015 | residential segregation measures; % poverty; % unemployed; income; % no high school; % disabled; % single-parent; % racial/ethnic minority; % speak English less than well; % multiunit structures; % mobile homes; % crowding; % no vehicle; % group quarters; % uninsured | % ages 65+ ; % less than 18 years old (also modeled as predictors) | negative binomial regression |
| Gabriel et al. (2021) | opioid overdose of any intent | 2009–2018 | % of 1860 population enslaved; number of lynchings (1865–1950); presence of active KKK chapter (1964–1966) | all covariates from 1860: population; % farms smaller than 5- acres; land holding inequity; farm value per acre; acres of improved farmland; % of free Black residents | censored, Bayesian multilevel negative binomial regression |
| Griffith et al. (2021) | opioid-related deaths | 2006-2013 | per capita oxycodone/ hydrocodone pill shipment volume | sex, age, and racial composition; cancer mortality rate; income; unemployment; % with 4+ years of college; MDs per population; specialist MDs per population; nurse practitioner prescribing authority; inpatient days per capita; outpatient visits per capita; % uninsured; % on Medicare; % dual-eligible (Medicare & Medicaid); number of hospices; short-term general hospital beds; long-term hospital beds; hospital-based | adjusted linear regression model with county-level fixed effects, weighted by county population |

Table 2 (continued)

| Author (year) | Outcome Measure | Years for Outcome Measure | County-Level Predictors | Controls/Covariates | Analysis |
|---|--|---|---|---|--|
| Grigoras et al. (2018) | opioid-involved overdose deaths | 2010–2014 pooled | opioid prescription rate to Medicare enrollees; % in poverty; racial composition; % male; age composition; population density; number of Medicare-enrolled opioid prescriber physicians per | nursing home beds; early Medicaid expansion (state-level) none | multivariable linear regression with spatial regression analyses |
| Gunn et al. (2018) Hadland et al. (2019) | unintentional poisoning deaths involving: opioids; heroin prescription opioid-related overdose deaths of any intent | 1999–2016 Aug 1, 2014- December 31, 2016 | rurality; opioid prescriptions; HIDTA designation pharmaceutical marketing of opioids to physicians; opioid prescribing rates | none % male; mean age; main race/ ethnicity; unemployment; income; poverty; % without education beyond high school; Gini index for income inequality: metropolitan status | logistic regression with random county effects negative binomial regression; mediation analysis |
| Haley et al. (2019) | opioid-related deaths | 2012–2014 | geographic access to: 1. FQHCs; 2. opioid treatment programs; 3. Buprenorphine providers | racial composition; % below poverty level; % without high school diploma | spatial error model regression |
| Hall (2022) | opioid overdose deaths | 2020 | presence of telemedicine; number of telemedicine service points | none | independent samples t-test and simple linear regression |
| Henry et al. (2021) | opioid overdose deaths of any intent: overall and for prescription opioids | 2011–2018 | presence of opioid safety coalition; initiative support of coalition | presence of substance use disorder treatment center "hub," number of "spokes" in treatment center "hub" per population | interrupted time series; mixed effects negative binomial regression with random intercepts for county and an order 1 autoregressive correlation structure |
| Hollingsworth et al. (2017) | drug overdose deaths of any intent, involving: any drug; and opioids | 1999–2014 | unemployment rate | not used in models with county and year fixed effects | regression models with county and year fixed effects |
| Hsu and Kovacs (2021) | opioid overdose deaths of any intent, overall and involving: prescription opioids; heroin; and synthetic opioids | 2015–2018 | number of cannabis dispensaries: overall, and medical and recreational dispensaries specifically | legalization of recreational cannabis sales; naloxone access laws; pain management laws; pharmacist patient identification laws; population size; unemployment; disability; income; % white; % bachelor's degree; % male; yearly GDP | panel regression |
| Jones et al. (2020) | unintentional opioid-related deaths | 2010–2012; 2015–2017 | per capita: opioid pills; buprenorphine prescriptions; EMS naloxone administrations; certified peer support specialists | none | Pearson product moment correlations |
| Kelly et al. (2021) | psychoactive drug overdose deaths of any intent | 1999–2016 | metropolitan status; unemployment; poverty; income; % with a Bachelor's degree; % foreign-born; % female-headed households; % Black; % Hispanic; state-level per-capita spending on education, public welfare, hospitals, and health; state-level PDMPs, Expanded Naloxone Access, Good Samaritan Laws, pain clinic prescribing restrictions, Medical Marijuana Laws | none | multilevel linear regression models |
| Kerry et al. (2016) | drug poisoning deaths of any intent | 2006–2010 | income; delayed physician use due to cost; population density; poverty; unemployment; median elevation; binge drinking; depression in Medicare enrollees; mentally unhealthy days; physically unhealthy days; smoking prevalence; suicide; average family size; % families; % Hispanic; % White; Latter-day Saints rate; YPLL | none | multiple linear regression and geographically weighted regression, including only variables strongly correlated with the outcome |
| Kerry et al. (2019) | drug poisoning deaths of any intent | 2006–2010 | average family size; % family households; Latter-day Saints rate; % White; % Hispanic; income; delayed physician use due to cost; population density; poverty; unemployment; elevation; binge drinking; depression in Medicare enrollees; mentally unhealthy | none | spatial profile regression |

Table 2 (continued)

| Author (year) | Outcome Measure | Years for Outcome Measure | County-Level Predictors | Controls/Covariates | Analysis |
|----------------------------|--|---------------------------------|--|---|---|
| Kline et al. (2019) | opioid poisoning deaths | 2013–2015 | days; physically unhealthy days; smoking; suicide, YPLL health professional shortage area; disability rate; unemployment; % white; median age; % with Bachelor's degree; % single female | none | Bayesian Poisson regression models |
| Kurani et al. (2020) | Bayesian smoothed drug poisoning mortality rates for overdoses of any intent, any | 2012–2017 | households Area Deprivation Index (includes 17 measures related to poverty, education, housing, and employment) | age composition, % White, % male, year | negative binomial regression |
| Langabeer et al. (2020) | opioid-involved overdose deaths of accidental or undetermined intent | 2016–2017 | employment) tobacco use rate; food insecurity rate; HIV prevalence; adult uninsurance rate; unemployment rate; % rural; % ages 65+; % NH White: % female | none | multivariate linear regression models with variables with a significant correlation to the outcome; stepwise backward selection |
| Lee et al. (2019) | opioid-related overdose deaths of any intent, any opioid | 2016 | opioid-related industry payments to providers | none | Spearman rank correlation |
| Lin et al. (2020) | drug overdose deaths of any intent, involving: any drug; any opioid; and not involving opioids | 2003–2014 | healthcare intensity index (ratio of local versus average national healthcare utilization rates) | % female; % racial/ethnic minority; % age; education; % foreign-born; % female-headed households; hospital beds and physicians per 1000 population; state PDMPs; medical/ recreational marijuana legalization; rural-urban continuum code; income; unemployment; poverty | regression with state and year fixed effects (linear models and dose-response models with quintiles) |
| Lister et al. | opioid overdose deaths of any intent involving any opioid | 2013-2017 | urban/rural classification | none | Mann-Whitney U tests |
| Marotta et al. (2019) | drug overdose deaths of any intent involving: any drug; any opioid; heroin; opioid pain relievers; methadone; synthetic opioids | 2013–2015 | prescriptions for opioid analgesics; racial/ethnic composition | median income | spatial regression modeling |
| Monnat (2018) | drug-related deaths (drug poisoning of any intent, drug- induced disease, drugs in the blood, mental/ behavioral disorders due to drugs) | 2006–2015 | indices and measures regarding: economic distress; housing distress; industry classification; family distress; residents living in different county five years earlier; social capital; healthcare environment | metropolitan status; racial composition; age composition; % military/veterans | generalized spatial two-stage least squares autoregressive models |
| Monnat (2019) | drug-related deaths | 2014–2016 | metropolitan status; economic distress index; family distress index; persistent population loss; industry classification; opioid prescriptions; opioid prescribing in neighboring counties; [exposure to fentanyl (state-level)] | age composition; racial composition; % military/ veterans | linear random effects regression models |
| Monnat et al. (2019) | drug-related deaths; opioid deaths; heroin deaths; prescription opioid deaths; synthetic opioids deaths; multiple cause (2+ opioids) deaths | 2002–2004; 2014–2016 | racial and age composition; % veterans; % moved into county in last 5 years; population density; nonmetropolitan status; % not working; % no 4-year college degree; ratio of federal to county median household income; poverty rate; public assistance rate; Thiel's L [income inequality]; Gini coefficient [income inequality]; % separated or divorced; % single-parent families; % vacant housing units; % renter- occupied housing units; occupational composition; industry composition; opioid prescribing | % NH White; % age 65+ ; population density (also modeled as predictors) | multilevel negative binomial regression |
| Nosrati et al. (2019) | drug use disorder-related deaths | 1983–2014 | jail and prison admission rates; median household income | violent crime rate; racial/ethnic composition; opioid prescription rate; all-cause mortality rate | two-way fixed-effects panel regression |
| Ombach et al. (2019) | prescription opioid overdose deaths of accidental or undetermined intent | 2006–2016 | mean elevation | urban-rural classification; opioid prescribing rates | logistic regression using Williams method |
| | drug overdose deaths | 1990–2013 | | | difference-in-difference (continued on next page) |

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Table 2 (continued)

| Table 2 (continued) | | | | | |
|--------------------------------|--|---------------------------------|---|--|---|
| Author (year) | Outcome Measure | Years for Outcome Measure | County-Level Predictors | Controls/Covariates | Analysis |
| Pierce and Schott (2020) | | | trade liberalization (exposure to Permanent Normal Trade Relations) | average US import normal trade relations tariff for county- produced goods; average exposure to phasing out of global Multi-Fiber Arrangement; average changes in Chinese import tariffs/ domestic production subsidies; income; % without college degree; % veteran; % foreign- born; manufacturing iob share | |
| Rangachari et al. (2022) | opioid use deaths | 2009-2018 | % female; median age; % Black; % White; % Hispanic; % Asian; % with bachelor's degree; median income; % in poverty; % unemployed; % institutionalized; % veterans; % in armed forces; number of FQHCs; number of rural health clinics; number of households; % of occupied housing units rented; % of housing units that are mobile homes | none | regression analysis with county, year, and state-by- year fixed effects |
| Romeiser et al. | prescription opioid overdose | 2013-2015 | opioid prescriptions | none | Poisson regression |
| Rowe et al. (2022) | opioid-related deaths of any intent | 2010–2018 | buprenorphine NP/PA waiver capacity; buprenorphine MD/DO- 275 waiver capacity | buprenorphine prescribing capacity rate; unemployment; % insured; non-buprenorphine opioid prescription rate; proportion of opioid-related deaths involving fentanyl | Poisson regression models with county and quarter level fixed effects |
| Rudolph et al. (2020) | drug overdose deaths of any intent | 2015 | unemployment rate; labor-force nonparticipation rate; relative unemployment rate (ratio of rate in subgroup vs. rate in overall county); relative labor-force nonparticipation rate (ratio of rate in subgroup vs. rate in overall county) | % of county population in the racial/ethnic sub-group; % of population in gender subgroup; % ages 0–19; % ages > 64 years; ratio of % 0–19 in subgroup to % 0–19 in overall population; ratio of % > 64-year in subgroup to % > 64 in overall percentage; drug overdose deaths in 2000 and 2009 | longitudinal-g computation; negative binomial regression |
| Ruhm (2019) | drug poisoning deaths of any intent, involving: any drug; opioid analgesics; illicit opioids (heroin and synthetic opioids) | 1999–2015 | unemployment rate; poverty rate; median household income; median home prices; exposure to imports | % female; racial composition; age composition; % with some college; % female-headed households; % foreign-born; number of hospital beds; number of physicians per 1000; rural-urban continuum; PDMP (state-level); medical and recreational cannabis legalization (state-level) | fixed effects regression models |
| Sauber-Schatz et al. (2013) | drug-caused deaths; opioid- caused deaths; oxycodone- caused deaths | 2009 | pain clinics; opioid pills distributed to pharmacies/ physicians; oxycodone pills distributed to pharmacies/ physicians | age composition; racial composition; % male | negative binomial regression |
| Sawyer et al. (2021) | opioid-involved overdose deaths | 2017 | opioid-related emergency department visits; opioid-related arrests; chronic HCV cases; opioid prescriptions; median income | % female-led households; unemployment; Gini index; % without high school diploma; % with disability; % NH Black; opioid use disorder services rate; % internet access | bivariate negative binomial regression models |
| Smith (2020) | prescription opioid overdose deaths of any intent | 1999–2014 | existence of medical marijuana dispensary | % age; % White; % male; unemployment, average weekly wages; PDMP (state-level) | difference-in-difference |
| Sun (2021) | opioid overdose mortality of any intent: overall and for prescription opioids, heroin, and synthetic opioids | 2005–2018 | rurality (index of relative rurality) | % ages 15–24; % ages 65+; % male; % White; % Black; % Hispanic; % with high school degree or higher; % in agriculture, forestry, fishing and hunting, mining, construction, and manufacturing; % divorced/separated; % with disability; diabetes prevalence; poverty; unemployment; | spatial panel regression, two- way fixed effects models |

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Table 2 (continued)

| Author (year) | Outcome Measure | Years for Outcome Measure | County-Level Predictors | Controls/Covariates | Analysis |
|---|---|---------------------------------|---|---|--|
| | | | | uninsurance; % with Medicaid; % veterans; % receiving public assistance; social capital index; number of active MDs, nurse practitioners, physician assistants, dentists, chiropractors; number of pain management doctors, surgeons, sports medicine doctors, oncologists | |
| Suriaga (2021) | opioid-related deaths of any intent | 2014–2018 | metro status; median income; % high school and higher; racial | none | generalized linear model |
| Swensen (2015) | drug-induced poisoning deaths | 1999–2008 | number of substance abuse treatment facilities | unemployment rate; firm births; per-capita income; racial | weighted regression models with county and year fixed |
| Tacheva and Ivanov (2021) | opioid overdose deaths of any intent | 2014–2016 | score for county: openness; conscientiousness; extroversion; agreeableness; and neuroticism, as assessed via linguistic analysis of Twitter posts | composition; age composition YPLL; low birth weight rate; % BMI> =30; access to healthy foods; % no leisure physical activity; % access to places for physical activity; % driving deaths with alcohol; STD rate; teen birth rate; uninsurance; discharges for ambulatory care sensitive conditions in Medicare enrollees; diabetic Medicare enrollees; diabetic Medicare enrollees; receiving HbA1c test; female Medicare enrollees with mammogram in two years; % postsecondary education; unemployment; child poverty; % children in single-parent households; % households with overcrowding, high costs, no kitchen, or no plumbing; % driving alone to work; % car commutes > 30 min; words in aggregated tweets | panel interval regression models with control function |
| Valentini and Jayawardhana (2019) | drug overdose deaths due to any drug except alcohol, and opioid overdose deaths | 1999–2015 | rural/non-rural status | age composition; racial composition; % male; income; poverty; unemployment; % uninsured; primary care physicians per 1000 population | multivariate linear regression |
| Venkataramani et al. (2020) | opioid overdose deaths of any intent; in secondary analyses, drug overdoses overall, prescription opioids, and "illicit opioids" (opium and heroin) | 1999–2016 | auto assembly plant closures | none in main analysis | difference-in-difference, regression models with county and year fixed effects |
| Whitley et al. (2022) | drug overdose deaths of all intents | 2013–2020 | urine drug test positivity for: heroin; other opioids (natural/ semisynthetic); synthetic opioids; cocaine: and methamphetamine | none | Poisson regression models, county fixed effects |
| Wu and Evangelist (2022) | opioid overdose deaths of any intent | 1999–2012 | mass layoff rate; non-layoff unemployment rate | state-level maximum unemployment insurance benefit; GDP; income; poverty; population; unemployment; unemployment insurance weekly wage; SNAP recipients; TANF recipients; Medicaid enrollees; state earned income tax credit rate; minimum wage | difference-in-difference |
| Yang et al. (2021) | opioid-related deaths of any intent | 2015–2017 | unemployment rate; social capital index; social isolation index; smoking prevalence | % White; % Black; % male; median age; dependency ratio; % uninsured; MDs per capita; opioid prescribing rate; net migration rate; affluence index; metropolitan status | path analysis |
| Zemore et al. (2022) | drug-related mortality | 2008–2017 | on- versus off- border county designation | age/ gender / racial composition; education; poverty; employment rate; urbanicity; HIDTA status | spatial lag models |
| Zhu et al. (2022) | opioid overdose mortality of any intent | 2014–2018 (pooled) | population size; population density; % Hispanic; % Black; % Asian; % Native Hawaiian or Other | none | generalized additive models with a thin plate spline term of spatial coordinates (continued on next page) |

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Table 2 (continued)

| Author (year) | Outcome Measure | Years for Outcome Measure | County-Level Predictors | Controls/Covariates | Analysis |
|---------------|---|---------------------------------|--|--|------------------------------------|
| | | | Pacific Islander; % other race; % two or more races; % male; % 25–64; % 65+; % high school; % some college or higher; income; % unemployed; physicians per population; healthcare professional shortage area score | | |
| Salemi (2017) | smoothed rates of drug overdose mortality (any intent, any drug) classified as low, moderate, and high | 1999–2014 | social capital index (based on civic associations/non-profits, voting rate, Census response rate, and non-profit organizations) | income; poverty; racial/ethnic composition; rurality; specialized substance use treatment centers per capita; opioid prescriptions for Medicare enrollees; year | multinomiai logistic regression |
| Zoorob (2018) | smoothed rates of drug overdose mortality (any intent/ drug); low, medium, high, very high | 2013–2015 | percentage of all Medicare Part D claims that are for opioids and benzodiazepines | income; poverty; education; race/ethnicity; urban-rural index | multinomial logistic regression |

Abbreviations. 2SRI, Two-stage Residual Inclusion; BMI, Body Mass Index; EMS, Emergency Medical Services; FQHC, Federally Qualifying Health Center; GDP, Gross Domestic Product; HIDTA, High Intensity Drug Trafficking Area; HCV, hepatitis C virus; HIV, Human Immunodeficiency Virus; KKK, Ku Klux Klan; MD, Medical Doctor; NP, Nurse Practitioner; OLS, Ordinary Least Squares; PA, Physician Assistant; PDMP, Prescription Drug Monitoring Program; SNAP, Supplemental Nutrition Assistance Program; TANF, Temporary Assistance for Needy Families; YPLL, Years of Potential Life Lost

healthcare systems and access, and health indicators), *economic factors* (e.g., economic health, employment, industries, income and poverty, housing), *social factors* (e.g., social capital, education, religion, family), and *demographic or geographic characteristics* (e.g., racial/ethnic, sex, and age composition, rurality, elevation). While some studies focused on a single predictor (e.g., Ombach et al., 2019; Romeiser et al., 2019; Valentini and Jayawardhana, 2019), others included 30 or more predictors (Frankenfeld and Leslie, 2019; Monnat et al., 2019). Studies also varied widely in the number and type of controls/covariates included in analyses, ranging from no controls/covariates (e.g., Hall, 2022; Jones et al., 2020; Lee et al., 2019; Lister et al., 2020), to more than 20 controls/covariates (e.g., Sun, 2021; Tacheva and Ivanov, 2021).

3.3. Quality Assessment

Results of the quality assessment are provided in Supplemental Table 5. All studies included a clearly defined research question, population, exposure, and outcome. While 89.3% of studies provided sample size information/justification, lower proportions mentioned adjusting for specific controls/covariates (62.5%) or evidenced temporal precedence (41.1%).

Supplemental Table 6 summarizes information about the extent to which included studies addressed the specific risk of bias due to underreporting of drug types on death certificates (Jones et al., 2019). Of the 56 studies in the review, 38 (67.9%) included analyses with overdose outcomes related to specific drug types, which are subject to underreporting. Among these 38 studies examining specific drug involvement in overdose deaths, five studies (Hollingsworth et al., 2017; Lin et al., 2020; Ruhm, 2019; Smith, 2020; Wu and Evangelist, 2022) reported using imputation to correct for opioid underreporting, while two other studies mentioned other corrections (e.g., additionally examining deaths without opioids reported [Monnat et al., 2019] or adding supplemental analyses with drug overdoses overall [Venkataramani et al., 2020]).

3.4. Results from the Included Studies

Table 3 summarizes main findings from each study, while Figs. 2–4 provide an overview of the directions of significant associations between county-level predictors and drug overdose mortality outcomes. Health-related predictors are depicted in Fig. 2, economic and social predictors in Fig. 3, and demographic and geographic predictors in Fig. 4. In each figure, the + symbol signifies a significant positive association

between the listed predictor and the drug overdose mortality outcome examined in the particular study, while the – symbol signifies a significant negative association, and the + /- symbol is used when both significant positive and negative associations were reported in the study across different models or different subpopulations. Most of these associations are conditional on other variables included in the regression model (Table 2 presents the variables included in each study's statistical analysis/analyses). Nearly all frequentist approaches reported statistically significant associations with a threshold of p < 0.05, while Bayesian approaches reported significant associations via 90% (e.g., Gabriel et al., 2021) or 95% intervals (e.g., Congdon, 2020).

With respect to health-related predictors (Fig. 2), various indicators of poor physical/mental health (disability rate, hepatitis C virus cases, prevalence of human immunodeficiency virus, physically unhealthy days, mentally unhealthy days, depression in Medicare beneficiaries, suicide rate, and years of potential life lost; Cordes, 2018; Frankenfeld and Leslie, 2019; Kerry et al., 2016, 2019; Langabeer et al., 2020; Sawyer et al., 2021) were positively associated with drug overdose mortality outcomes. Associations between substance use-related indicators (e.g., smoking; Kerry et al., 2016; Langabeer et al., 2020; Yang et al., 2021) and drug overdose mortality outcomes were mixed, however. Measures related to opioid prescribing (whether opioid prescribing rates, pill shipments, pain clinics, or opioid marketing/ industry payments; Blake-Gonzalez et al., 2021; Griffith et al., 2021; Grigoras et al., 2018; Gunn et al., 2018; Hadland et al., 2019; Lee et al., 2019; Marotta et al., 2019; Monnat, 2019; Monnat et al., 2019; Sauber-Schatz et al., 2013; Sawyer et al., 2021) were consistently positively associated with drug overdose mortality outcomes, with one exception (Romeiser et al., 2019). Findings regarding healthcare systems and access were inconsistent; for example, uninsurance rates were negatively associated with drug overdose mortality outcomes in two studies (Langabeer et al., 2020; Blake-Gonzalez et al., 2021) while both positive and negative findings were reported in another study (Frankenfeld and Leslie, 2019). Although mixed results were observed regarding the healthcare system in general (e.g., healthcare professional shortage area, physicians per capita; Grigoras et al., 2018; Kline et al., 2019; Zhu et al., 2022), measures related to substance use disorder treatment access were more consistently associated with lower drug overdose mortality outcomes (Haley et al., 2019; Jones et al., 2020; Swensen, 2015).

As presented in Fig. 3, associations between economic indicators and drug overdose mortality outcomes were relatively mixed. Although economic distress and economic decline were positively associated with drug-related mortality (Monnat, 2018, 2019; Ruhm, 2019), Ruhm

Table 3

Findings regarding associations between county-level predictors and drugrelated mortality outcomes, as reported by the 56 articles included in the systematic review.

| Author (year) | Main Findings | |
|----------------------------------|---|-------------------------|
| Betz and Jones (2018) | Associations between employment growth or wage growth and opioid overdose mortality vary by tier of industry gender race and metro/popmetro counties | Gunn et al. (20 |
| | Negative associations between lower-tier industry employment growth or wage growth and opioid overdose mortality; positive association between top- tier industry employment growth and opioid overdose | Hadland et al. |
| Blake-Gonzalez et al. (2021) | mortality. In Virginia: opioid prescribing rate, mining employment share, % insured, incarceration rate, mean commute time, and unemployment rate positively associated with drug overdose mortality; market share of five largest pharmacies, median income, and | Haley et al. (20 |
| Bradford and Bradford | population density negatively associated with drug overdose mortality. Positive association between eviction rate and | Hall (2022) |
| (2020) | substance-related mortality (for all opioids, prescription opioids, synthetic opioids, heroin, benzodiazepines, and alcohol poisoning) in all counties; these positive associations were also observed in urban counties (accent for prescription opioids), but | Henry et al. (20 |
| Bruch et al. (2021) | not consistently observed in suburban or rural counties. Nonsignificant association between border wall | Hollingsworth (2017) |
| Congdon (2020) | construction and drug overdose mortality. Association with drug overdose mortality was: positive for unemployment rate, unemployment rate growth, and projet correction proceims for Index of | Hsu and Kovac |
| | and racial segregation; negative for index of Concentration at Extremes (income inequality), social capital, and rurality. In analyses by geographic division: income inequality and rurality were negatively associated with drug overdose mortality across all divisions, while the significance of associations between drug mortality and unemployment, social capital, or segregation varied between geographic divisions. | |
| Cordes (2018) | In North Carolina: Association with prescription opioid mortality was: negative for percent Black and percent Hispanic, positive for percent disability. Association with heroin mortality was: negative for rurality and | Jones et al. (20 |
| Dean and Kimmel (2019) | positive for percent with college education. Positive association between trade-related job loss and opioid-related mortality; association strengthened | Kelly et al. (20 |
| Frankenfeld and Leslie (2019) | When rentary is present in the county's heroin suppy. Overall, positive associations between overdose mortality rates and multiple residential segregation measures, as well as % unemployed, income, % without high school diploma, % civilian disabled, % single- parent household, % without a vehicle. Overall, negative associations between overdose mortality and: % in poverty, % racial/ethnic minority, % speaking English less than well, % multiunit structures, % mobile home housing, % group quarters, and % crowded housing. Association with uninsurance rate varied (positive/ negative) based on model. Associations with overdose mortality differed by race/ethnicity (unemployment positive for White, negative for Hispanic; income positive for Black, negative for Black; percent uninsured null for White, negative for Black and Hispanic). Association between racial diversity and | Kerry et al. (20 |
| Gabriel et al. (2021) | overdose mortality was null for White population and negative for Hispanic and Black populations. In the southern states examined: proportion enslaved (in 1860) was negatively associated with contemporary | Kerry et al. (20 |
| Griffith et al. (2021) | opioid mortality in the Non-Hispanic White population. Positive association between per capita oxycodone/ hydrocodone pill shipment volume and opioid-related | |
| Grigoras et al. (2018) | deaths. Positive associations between opioid overdose mortality and: rates of opioids prescribed (overall and by emergency medicine, family medicine, internal medicine, physician assistants), poverty rate, and | Kline et al. (20 |

| Author (year) | Main Findings |
|-------------------------|--|
| | White population share; negative association for |
| | number of opioid prescribing physicians per |
| | population. |
| Gunn et al. (2018) | In North Carolina: Positive association between opioi |
| | prescriptions per capita and unintentional opioid |
| | unintentional heroin mortality. |
| Hadland et al. (2019) | Positive association between all three measures of |
| | pharmaceutical industry opioid marketing and |
| | prescription opioid overdose mortality; prescribing |
| Islam at al. (2010) | rates partially mediated this relationship. |
| Taley et al. (2019) | onioid mortality rates and geographic access to a) |
| | opioid treatment programs and b) Federally Qualified |
| | Health Centers. Negative significant association |
| | between buprenorphine provider concentration and |
| | opioid mortality in 2014 but not 2012. |
| Hall (2022) | In Texas: no significant association between the |
| | presence of telemedicine, or number of telemedicine |
| Jenry et al. (2021) | In California: presence of a county opioid safety |
| ionij et un (2021) | coalition was negatively associated with opioid |
| | overdose mortality and prescription opioid overdose |
| | mortality. |
| Iollingsworth et al. | Positive association between unemployment rate and |
| (2017) | opioid-related deaths or drug-related deaths in overa |
| | population and White population. In Black population |
| | opioid deaths |
| Isu and Koyacs (2021) | In population ages $21+$ within states with legalized |
| | cannabis dispensaries: the number of cannabis |
| | dispensaries (and the number of medical dispensaries |
| | specifically) was negatively associated with overdose |
| | deaths due to any opioid, prescription opioids, heroin |
| | or synthetic opioids. For recreational dispensaries |
| | specifically, negative associations with overdose |
| | overall and synthetic opioids, but not prescription |
| | opioids or heroin. |
| lones et al. (2020) | In North Carolina: opioid-related mortality positively |
| | associated with change in number of EMS naloxone |
| | administrations and negatively associated with chang |
| r 11 1 (2001) | in buprenorphine prescription rates. |
| (elly et al. (2021) | Proportion of female-headed households negatively |
| | overdose mortality |
| (2016) Kerry et al. | In US overall: positive associations between drug |
| | poisoning death rate and: median household income; |
| | delaying use of physician due to cost; unemployment |
| | elevation; depression in Medicare beneficiaries; |
| | mentally unhealthy days; physically unhealthy days; |
| | suicide rate; percent White; and years of potential life |
| | lost; negative associations with binge drinking, |
| | positive associations between drug poisoning death |
| | and: depression in Medicare beneficiaries: mentally |
| | unhealthy days; years of potential life lost; percent |
| | Hispanic; and Latter-Day Saints rate; negative for |
| | population density; poverty rate; and percent family |
| | households. In Utah: positive associations between |
| | drug poisoning death and: population density; |
| | unemployment; and years of potential life lost; negativ |
| (erry et al. (2019) | In all western states: positive associations between dru |
| (LILY (L (LL (LL (LL)) | overdose mortality and Latter-Day Saints rate. |
| | elevation, and suicide rate; negative associations with |
| | median household income, population density, and |
| | unemployment. In "high-risk" western states: positive |
| | association between drug overdose mortality and |
| | percent Hispanic, poverty, unemployment, mentally |
| | unnealthy days, physically unhealthy days, years of |
| | potential life lost; negative associations with percent |
| | White median household income population density |
| Kline et al. (2019) | In Ohio: Association with opioid poisoning mortality |
| | was: positive for percent of single-female households: |
| | r r r r r r r r r r r r r r r r r r r |

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| Author (year) | Main Findings | |
|---------------------------------------|---|--|
| · · · · · · · · · · · · · · · · · · · | and negative for location in a Health Destactional | |
| | and negative for location in a Health Professional Shortage Area. | |
| Kurani et al. (2020) | Positive association between Area Deprivation Index | |
| | quintile and drug poisoning mortality. | |
| Langabeer et al. (2020) | Positive associations between opioid mortality and | |
| | White, and percent female: negative associations | |
| | between opioid mortality and food insecurity rate and | |
| | uninsurance rate. | |
| Lee et al. (2019) | Positive association between opioid-related industry | |
| Lin et al. (2020) | payments and opioid-related mortality. | |
| Liii et al. (2020) | opioid-involved mortality in the overall population, as | |
| | well as both sexes, all educational levels, ages 20-44 | |
| | and 45–64, White and Black (but not Hispanic or other | |
| | races); no association with non-opioid overdose | |
| Lister at 21 (2020) | mortality. In Michigan: Urban (vs. rural) counties had higher | |
| Lister et al. (2020) | opioid overdose death rates. | |
| Marotta et al. (2019) | In New York state: Positive association between opioid | |
| | prescriptions and overdose mortality from any opioid, | |
| | opioid pain relievers, and synthetic opioids. Positive | |
| | association between percent Black and heroin overdose | |
| Monnat (2018) | Association with drug-related mortality was: positive | |
| | for economic distress, rental stress, labor market | |
| | dependence on mining, and family distress; negative | |
| | for labor market dependence on public sector, | |
| | proportion of residents living in other counties five | |
| Monnat (2019) | years prior, and number of religious establishments. Positive associations between NH White drug-related | |
| | mortality rate and economic distress, family distress. | |
| | mining and service industry, persistent population loss, | |
| | and opioid prescribing rates; negative association with | |
| | rurality (vs. large metro status) and manufacturing or | |
| | public sector economic dependence. Some associations | |
| | labor markets. | |
| Monnat et al. (2019) | Positive association with drug-related mortality for | |
| | percent NH White, percent 65+, population density, | |
| | percent not working, percent with no 4-year college; | |
| | ratio of federal to county median household income, | |
| | coefficient, percent separated/divorced, percent single- | |
| | parent family, percent vacant housing units, opioid | |
| | prescribing rate, occupational percent in personal | |
| | services, and industry percent in health, retail/personal | |
| | services/tood/accommodations, mining, and public | |
| | mortality for percent NH Black, percent moved into | |
| | county in last 5 years, occupational percent in | |
| | executive/managerial or farming/fishing/forestry, and | |
| | industry percent in agriculture/fishing/forestry, | |
| | pusiness/professional, finance/insurance/real estate, | |
| Nosrati et al. (2019) | Positive association between jail and prison admission | |
| | rates and drug use disorder mortality rates; negative | |
| | association between household income and drug use | |
| 0 | disorder mortality. | |
| Umbach et al. (2019) | Positive association between mean elevation and | |
| | population and male and female populations. | |
| Pierce and Schott (2020) | Positive association between exposure to Permanent | |
| | Normal Trade Relations and drug overdose mortality | |
| | rate in the overall and White populations (both males | |
| | and females) but not for any other races. | |
| Kangachari et al. (2022) | Negative association between opioid use mortality and: | |
| | percent veterans. | |
| Romeiser et al. (2019) | In New York state: negative association (described by | |
| | authors as "small" and "essentially null") between | |
| | opioid prescription rate and prescription opioid | |
| Devue et al. (0000) | mortality. | |
| kowe et al. (2022) | IN California: no significant associations between NP/ | |
| | opioid-related mortality. | |
| | r | |

| Author (year) | Main Findings |
|--|---|
| Rudolph et al. (2020) | Generally positive associations between overall and relative unemployment rates and drug overdose mortality both in the short and long term for males across racial/ethnic groups. For overall and relative labor-force nonparticipation rates, associations with drug overdose mortality differed by gender, race/ ethnicity, and short long term timeframe. |
| Ruhm (2019) | Positive associations between economic decline measures and overdose deaths due to all drugs, opioid analgesics, or illicit opioids were substantially attenuated or became non-significant after adjusting for covariates |
| Sauber-Schatz et al. (2013) | In Florida: Positive association between rates of opioid or oxycodone pills distributed and overdose deaths due to any drugs, opioids, or oxycodone. Positive association between pain clinic rate density rate and oxycodone overdose deaths only in the unadjusted |
| Sawyer et al. (2021) | model but not adjusted model. In Indiana: opioid-related emergency department visit rate, opioid-related arrest rate, opioid prescription rate, and chronic HCV cases positively associated with opioid-involved overdose deaths (in bivariate models). |
| Smith (2020) | Negative association between opening of a medical marijuana dispensary and prescription opioid-related deaths- for adults overall and for males, females, and NH White males. |
| Sun (2021) | Index of relative rurality was negatively associated with opioid-related mortality (for all opioids, prescription opioids, heroin, and synthetic opioids) in the US overall. Index of relative rurality was positively associated with opioid-related mortality in urban counties; no significant relationship in rural counties. |
| Suriaga (2021) | In Florida: % with high school degree or higher positively associated with opioid death rate in older adults (ages 65+). |
| Swensen (2015) | Negative association between substance use disorder treatment facilities and drug-induced mortality rate, for facilities overall and for outpatient facilities; Negative associations with number of facilities observed across ages 10+, all racial groups, male and female populations and in large urban and medium/small urban (but not rural) and medium and high income (but not low incomp) counties |
| Tacheva and Ivanov (2021) | Positive associations between opioid overdose mortality and county Twitter-based scores of: conscientiousness, extraversion, and neuroticism. |
| Valentini and Jayawardhana (2019) Venkataramani et al. (2020) | In Georgia: Negative association between rurality (vs. urban) and drug overdose mortality rate. Positive association between automotive assembly plant closures and opioid overdose mortality in overall population and in NH White men ages 18–34 and 35–65 and NH White women ages 18–34 |
| Whitley et al. (2022) | In Ohio: positive bivariate associations between drug overdose mortality rates and urine test positivity rates for cocaine, synthetic opioids, and methamphetamine; negative bivariate associations between drug overdose mortality rates and urine test positivity rates for heroin and other (natural/semisynthetic) opioids. (Associations for methamphetamine and heroin nonsignificant in model that included positivity rates for all other drugs.) |
| Wu and Evangelist (2022) | Nonsignificant association between (a) mass layoff rate or (b) non-layoff unemployment rate and opioid- related mortality. However, state-level unemployment insurance benefit level moderated the association, for the overall population (ages 25–54) and for men, uponen White individuels across 25–24 and for men, |
| Yang et al. (2021) | women, winter individuals, ages 25–34, and 45–54. Unemployment was positively associated with opioid- related mortality. In path analysis, smoking was positively associated with opioid-related mortality; social capital was negatively associated with opioid- related mortality. Social capital, social isolation, and smoking were all significant mediators of the association between unemployment and opioid-related mortality. |

Zemore et al. (2022)

Table 3 (continued)

| Author (year) | Main Findings |
|---------------------------|--|
| Zhu et al. (2022) | In US-Mexico border states: county on-border (vs. non- border) status was not significantly associated with drug-related mortality in bivariate model but significantly negatively associated with drug-related mortality when adjusting for covariates. In both rural and urban counties: opioid overdose mortality rates positively associated with % two or more races and % ages 25–64; negatively associated with % Black, % male, and median income. In urban counties only: positively associated with |
| Zooroh and Salemi (2017) | unemployment rate and physicians per capita. In rural counties only: positively associated with healthcare professional shortage area score. |
| 20010D and Saleini (2017) | and drug overdose mortality rates |
| Zoorob (2018) | Positive association between percent of Medicare Part |
| | D claims for benzodiazepines (or opioids) and drug |
| | overdose mortality, with interaction between |
| | benzodiazenine and opioid claim percentages |

Abbreviations: EMS, Emergency Medical Services; HCV, hepatitis C virus; HIV, human immunodeficiency virus; NH, Non-Hispanic; MD/DO, Medical Doctor/ Doctor of Osteopathic Medicine; NP/PA, Nurse Practitioner/ Physician Assistant; US, United States.

(2019) noted that positive associations with economic decline measures were substantially attenuated or non-significant after adjusting for confounding factors. Associations between unemployment rates and drug overdose mortality outcomes varied between and within studies and subpopulations, with three studies reporting both positive and negative associations (Frankenfeld and Leslie, 2019; Hollingsworth et al., 2017; Kerry et al., 2019) and eight studies reporting positive associations (Blake-Gonzalez et al., 2021; Congdon, 2020; Kerry et al., 2016; Monnat et al., 2019; Ruhm, 2019; Rudolph et al., 2020; Yang et al., 2021; and Zhu et al., 2022), including one study concluding that the association was mediated by social capital, social isolation, and smoking (Yang et al., 2021). While mining employment share (Blake--Gonzalez et al., 2021; Monnat, 2018, 2019; Monnat et al., 2019) and specific large-scale job losses (Dean and Kimmel, 2019; Venkataramani et al., 2020) were consistently associated with higher overdose mortality outcomes, associations between overdose mortality and other economic indicators were more nuanced; for example, lower-tier industry employment/wage growth was negatively associated with opioid mortality, yet top-tier industry employment growth was positively associated with opioid mortality (Betz and Jones, 2018), and associations between overdose mortality and overall or relative labor-force nonparticipation rates varied by time-frame, gender, and race/ethnicity (Rudolph et al., 2020).

Associations between measures of income/poverty and overdose mortality were also mixed. Six studies (Blake-Gonzalez et al., 2021;



Fig. 2. Health-related county-level predictors of drug overdose mortality outcomes. Notes: The + symbol indicates a statistically significant positive association reported between the specified predictor and the overdose mortality outcome examined in the particular study, the - symbol indicates a statistically significant negative association, and + /- signifies that both significant positive and negative associations were documented in the study (in different analyses or subpopulations). While some associations are bivariate, others are conditional on the controls or covariates included in the statistical models. *Abbreviations*: HCV, hepatitis C virus; HIV, human immunodeficiency virus; EMS, Emergency Medical Services.



Fig. 3. Economic and social county-level predictors of drug overdose mortality outcomes. Notes: The + symbol indicates a statistically significant positive association reported between the specified predictor and the overdose mortality outcome examined in the particular study, the - symbol indicates a statistically significant negative association, and + /- signifies that both significant positive and negative associations were documented in the study (in different analyses or sub-populations). While some associations are bivariate, others are conditional on the controls or covariates included in the statistical models.

Kerry et al., 2019; Nosrati et al., 2019; Rangachari et al., 2022; Ruhm, 2019; Zhu et al., 2022) reported negative associations between income measures and overdose mortality outcomes, while one study (Kerry et al., 2016) reported a positive association and another study reported both positive and negative associations (differing based on racial/ethnic group; Frankenfeld and Leslie, 2019). Similarly, the poverty rate was positively associated with overdose mortality outcomes in four studies (Grigoras et al., 2018; Kerry et al., 2019; Monnat et al., 2019; Ruhm, 2019) yet negatively associated in two other studies (Frankenfeld and

Leslie, 2019; Kerry et al., 2016). With respect to social characteristics, higher social capital (Congdon, 2020; Yang et al., 2021; Zoorob and Salemi, 2017) and proportions of family households (Kerry et al., 2016, 2019) were consistently associated with lower overdose mortality outcomes, while higher rates of incarceration (Blake-Gonzalez et al., 2021; Nosrati et al., 2019), family distress (Monnat, 2018, 2019), and single-parent families were associated with higher overdose mortality rates.

Finally, Fig. 4 depicts associations between overdose mortality



Fig. 4. Demographic and geographic county-level predictors of drug overdose mortality outcomes. Notes: The + symbol indicates a statistically significant positive association reported between the specified predictor and the overdose mortality outcome examined in the particular study, the - symbol indicates a statistically significant positive and negative associations were documented in the study (in different analyses or subpopulations). While some associations are bivariate, others are conditional on the controls or covariates included in the statistical models.

outcomes and demographic or geographic predictors. The associations between racial composition and overdose mortality outcomes varied between studies. Associations between population density or rurality and overdose mortality outcomes were also mixed; population density was negatively associated with drug overdose mortality in two studies (Blake-Gonzalez et al., 2021; Kerry et al., 2019), positively associated in another study (Monnat et al., 2019), and both positively and negatively associated in different geographic areas within another study (Kerry et al., 2016). Rurality was associated with reduced overdose mortality outcomes in six studies (Congdon, 2020; Cordes, 2018; Gunn et al., 2018; Lister et al., 2020; Monnat, 2019; Valentini and Jayawardhana, 2019) yet associated with increased drug overdose mortality in another study (Langabeer et al., 2020), and the index of relative rurality was negatively associated with opioid-related mortality in the US overall while positively associated with opioid-related mortality in urban counties (Sun, 2021).

4. Discussion

This systematic review qualitatively synthesized associations between US county-level predictors and drug overdose mortality outcomes across 56 published studies. County-level variation in drug overdose mortality appears to represent a relatively recent yet quickly expanding topic of research; although our systematic review search included publications beginning in 1990, no publications prior to 2013 met criteria for inclusion, and 91.1% of included studies had been published between 2018 and July 2022, with 80.4% of studies published between 2019 and July 2022. The recent growth in this research topic may potentially reflect the increasing severity of the drug overdose crisis (Hedegaard et al., 2021), the public availability of county-level overdose mortality data across multiple online platforms (e.g., Centers for Disease Control and Prevention, 2022; Robert Wood Johnson Foundation, 2022), and recent, prominent conceptualizations of the social and economic roots of the overdose crisis and "deaths of despair" (e.g., Case and

Deaton, 2017; Dasgupta et al., 2018).

Across studies in this systematic review, measures related to opioid prescribing (Blake-Gonzalez et al., 2021; Griffith et al., 2021; Grigoras et al., 2018; Gunn et al., 2018; Hadland et al., 2019; Lee et al., 2019; Marotta et al., 2019; Monnat, 2019; Monnat et al., 2019; Sauber-Schatz et al., 2013; Sawyer et al., 2021), disability/illness (Cordes, 2018; Frankenfeld and Leslie, 2019; Kerry et al., 2016, 2019; Langabeer et al., 2020; Sawyer et al., 2021), economic distress (Monnat, 2018, 2019), mining employment (Blake-Gonzalez et al., 2021; Monnat, 2018, 2019; Monnat et al., 2019), incarceration (Blake-Gonzalez et al., 2021; Nosrati et al., 2019), family distress (Monnat, 2018, 2019), and proportions of single-parent families (Frankenfeld and Leslie, 2019; Monnat et al., 2019) were generally consistently associated with increased drug overdose mortality outcomes across multiple studies. In contrast, measures related to cannabis dispensaries (Hsu and Kovacs, 2021; Smith, 2020), substance use treatment (Haley et al., 2019; Jones et al., 2020; Swensen, 2015), social capital (Congdon, 2020; Yang et al., 2021; Zoorob and Salemi, 2017), and proportions of family households (Kerry et al., 2016, 2019) were generally consistently associated with lower drug overdose mortality outcomes across multiple studies. Results regarding smoking/tobacco use, uninsurance, healthcare professional shortage status, physicians per capita, unemployment, income, poverty, education, racial composition, and rurality were less consistent, with both positive and negative associations documented. Findings in this systematic review were relatively less consistent than those reported in a recent systematic review of associations between socioeconomic marginalization and overdose at the individual level (Van Draanen et al., 2020).

The variation in results across studies in the review may plausibly be rooted in differences in study time periods, settings, measures, and methodology. First, the specific years of data and geographic areas examined may influence study findings, considering that the overdose crisis has varied widely across years and regions, both in terms of severity and the drug types frequently implicated in overdose deaths (Hedegaard et al., 2019, 2021). Second, predictor and outcome measures varied considerably across the studies included in the review. For instance, some studies examined drug overdose deaths involving any type of drugs, while other studies focused on overdose deaths attributed to specific drugs. Individual studies that examined multiple overdose outcomes (each involving a different type of drug) often yielded divergent results depending on drug type; for example, healthcare intensity was associated with opioid-involved mortality yet not with non-opioid drug mortality (Lin et al., 2020). As such, some of the variation in the results of studies included in the systematic review may be rooted in the variation in types of drug overdoses examined. Furthermore, studies examining overdose deaths involving specific drugs (e.g., opioids, heroin) may be subject to the bias introduced by county-level variation in the extent to which specific drugs are identified and recorded on death certificates (Jones et al., 2019), while this specific source of bias does not apply to studies examining drug overdose deaths overall. Finally, varying results across studies in this systematic review may also stem from the different methodologies employed, such as panel data versus cross sectional analyses, one year of data versus several years pooled, and inclusion of spatial and/or time lags. Study variation in addressing missing data (e.g., listwise deletion versus imputation) may also contribute to the mixed findings, considering that even within a single study, results differed between analyses using complete cases only and analyses with an imputed sample (e.g., Frankenfeld and Leslie, 2019). Furthermore, the number and type of covariates included in models varied widely across studies, and even within individual studies, results were sensitive to the inclusion of particular covariates (Ruhm, 2019; Sauber-Schatz et al., 2013; Whitley et al., 2022; Yang et al., 2021; Zemore et al., 2022). The potential relevance of which controls/covariates are modeled is underscored by analyses indicating that confounding factors nearly fully accounted for associations between economic decline proxies and overdose deaths due to various drug types (Ruhm, 2019) and that the association between unemployment and opioid-related mortality became non-significant after adjusting for social capital (Yang et al., 2021).

Beyond methodological differences, mixed findings between and within studies may also reflect the complex, often mediated and moderated pathways and multidirectional associations between healthrelated, social, economic, and demographic factors and overdose mortality outcomes. For instance, healthcare access may be accompanied both by increased access to prescription opioid pain relievers (which may increase overdose risk) and increased access to treatment for substance use disorders (which may reduce overdose risk; Kravitz-Wirtz et al., 2020). Findings from this systematic review also emphasize subpopulation variation in county-level risk factors for drug overdose mortality. Predictors of drug-related mortality differed in urban versus rural counties (Bradford and Bradford, 2020; Monnat, 2019; Swensen, 2015; Zhu et al., 2022), low versus medium/high income counties (Swensen, 2015), men versus women (Rudolph et al., 2020), and individuals of different age ranges (Lin et al., 2020). Variation between racial/ethnic groups was also observed for associations between drug mortality outcomes and employment or wage growth (Betz and Jones, 2018), unemployment (Hollingsworth et al., 2017; Frankenfeld and Leslie, 2019), labor force non-participation (Rudolph et al., 2020), per capita income, racial and unemployment diversity, percentage of crowded housing (Frankenfeld and Leslie, 2019), healthcare intensity (Lin et al., 2020), and trade policies (Pierce and Schott, 2020). Prior literature has suggested that racial/ethnic groups differ in terms of individual-level risk factors for substance use-related outcomes (Cano and Gelpí-Acosta, 2022; Shih et al., 2010; Stone et al., 2012), yet it is unclear if, or how, such individual-level factors or processes may translate into differences observed at the county level. When considering why associations between certain predictors and drug overdose mortality vary by racial/ethnic group, it may also be relevant to note that analyses often model predictors measured for the overall county population while the outcome is measured within one specific racial group, even though, when considering an example such as median income, the

median income in any given county overall may not necessarily reflect the median income within a racial/ethnic minority subpopulation in that county.

4.1. Limitations

Studies included in this systematic review are subject to limitations such as geographic variation in the completeness of drug reporting on death certificates (Boslett et al., 2020; Jones et al., 2019), missing data for counties with suppressed mortality data, and limited evidence of conceptual frameworks guiding variable selection; moreover, temporal precedence, sensitivity analyses, corrections for multiple testing, and methods accounting for spatial autocorrelation were not consistently implemented across all studies.

In this systematic review, the measure of agreement between screeners was only "moderate" for the first stage of screening (title/ abstract screening), yet this moderate level of initial agreement was addressed by the use of a third independent reviewer to screen all titles/ abstracts with discrepancies in votes; near perfect agreement between reviewers was observed at the full text review stage. Although the present review included studies published through July 19 2022, all but two studies (Hall, 2022; Whitley et al., 2022) used mortality data solely from years prior to 2020; as such, it is unclear to what extent findings from the review may apply to county-level drug overdose mortality in the era of COVID-19.

The wide variety of predictors in the included studies precluded a meta-analysis of the estimates; therefore, this systematic review focused on identifying significant county-level predictors of drug-related mortality without examining the strength of these associations. We extracted data solely regarding associations identified by study authors as statistically significant; therefore, the findings reported were subject to the arguably arbitrary (Wasserstein et al., 2019), albeit conventional, cut-off points selected by study authors for p values or credible intervals. The effect sizes of most findings were relatively modest, and a determination of whether findings should be considered clinically significant, as opposed to only statistically significant, was outside the review scope. Finally, due to the study's focus on county-level predictors, evidence regarding policies at the state-level (e.g., Prescription Drug Monitoring Programs, Medicaid expansion, cannabis legalization) was outside of the review scope, as was evidence regarding "deaths of despair" beyond drug overdoses (i.e., alcohol poisoning, suicide, or alcoholic liver disease).

4.2. Implications

The studies included in this systematic review examined a wide range of potential predictors of drug overdose mortality, yet predictors related to illicitly-manufactured fentanyl exposure were examined notably less frequently than predictors related to prescription opioid pain relievers. Of the 56 studies in the review, only two included a county-level measure related to fentanyl exposure (as a moderator in Dean and Kimmel, 2019; as a control in Rowe et al., 2022), and two other studies modeled fentanyl exposure at the state-level (Congdon, 2020; Monnat, 2019). Deaths involving synthetic opioids such as fentanyl have surpassed deaths involving prescription opioid pain relievers since 2016 (Hedegaard et al., 2021); as such, future research may benefit from including measures related to the presence of illicitly-manufactured fentanyl in the drug supply, which varies regionally (Hedegaard et al., 2019). Furthermore, as the US overdose crisis has continued to evolve, stimulants such as methamphetamine and cocaine have been increasingly involved in overdose deaths, often in combination with opioids (Hoots et al., 2020); considering that only one study in the review (Bradford and Bradford, 2020) reported examining psychostimulant-involved and cocaine-involved mortality, and only one study examined a stimulant-specific predictor (Whitley et al., 2022), county-level measures related to stimulants may also represent a

relevant future research direction.

Future research may also benefit from increased utilization of the imputation procedures researchers have developed to correct for the underreporting of opioid involvement on death certificates (Boslett et al., 2020; Ruhm, 2018), as only five studies in this review (Hollingsworth, 2017; Lin, 2020; Ruhm, 2019; Smith, 2020; Wu and Evangelist, 2022) specifically reported employing such imputation procedures even though 38 studies examined drug-specific outcomes that are subject to underreporting limitations. In consideration of the inconsistencies in results reflected in this review, future researchers may wish to build on the example of studies that utilized multiple model specifications and sensitivity analyses to evaluate robustness of findings (e.g., Hollingsworth et al., 2017; Swensen, 2015; Ruhm, 2019), as well as moderation analyses to determine the extent to which findings apply across different subpopulations (Frankenfeld and Leslie, 2019; Hollingsworth et al., 2017; Lin et al., 2020; Monnat, 2019; Rudolph et al., 2020; Smith, 2020; Venkataramani et al., 2020; Wu and Evangelist, 2022; Zhu et al., 2022) or vary by state-level policies (Wu and Evangelist, 2022). Finally, mediators and potential pathways for associations represent a promising step in expanding the knowledge base on county-level predictors of drug overdose mortality, as mediation analyses were utilized in relatively few (e.g., Hadland et al., 2019; Yang et al., 2021) of the 56 studies included in this systematic review.

5. Conclusion

Researchers diverge in the extent to which they attribute the US overdose crisis to supply-side factors versus demand-related measures such as socioeconomic vulnerabilities (Ciccarone, 2019; Dasgupta et al., 2018; Fischer et al., 2020; Ruhm, 2019). Based on the studies included in this review, both county-level supply- and demand-related measures are associated with overdose mortality rates, although county-level demand-related predictors have been examined more frequently, and with a greater variety of indicators, than county-level supply-side factors. This observation regarding county-level measures contrasts with the conclusions of a narrative review of primarily state-level policies (Maclean et al., 2022) which documented more published quasi-experimental studies of health impacts from supply-side policies (e.g., prescription drug monitoring programs, pain clinic management laws) than demand-side policies (e.g., doctor shopping laws, cannabis legalization). The findings of the present systematic review provide relatively mixed evidence regarding many county-level predictors of overdose mortality, several of which also vary between subpopulations. As such, findings support the need for additional research to elucidate the pathways by which the county context shapes residents' risk of fatal overdose, in order to translate data into contextual interventions to reverse the rising rates of overdose deaths across the US.

CRediT authorship contribution statement

Manuel Cano: Conceptualization, Investigation, Data curation, Writing – original draft. Sehun Oh: Conceptualization, Investigation, Writing – review & editing. Preston Osborn: Investigation, Data curation. Samson Olowolaju: Investigation, Data curation. Armando Sanchez: Investigation, Data curation, Yeonwoo Kim: Conceptualization, Methodology, Writing – review & editing. Alberto Cano Moreno: Investigation, Data curation.

Role of funding source

Nothing declared.

Conflict of interest

No conflict declared.

Appendix A. Supporting information

Supplementary data associated with this article can be found in the online version at doi:10.1016/j.drugalcdep.2022.109714.

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