



Assessing the Impacts of Differential Privacy on Public Health Surveillance at Varying Geographic Resolutions

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The findings and conclusions in this presentation are those of the authors and do not necessarily represent the official position of the Centers for Disease Control and Prevention (CDC).

National Center for Environmental Health
Division of Environmental Health Science and Practice



Q: What is differential privacy?

- Creates uncertainty about the identities of people behind the numbers and allows for measurement of privacy risk.
- Adds precisely calibrated “statistical noise” to data tables published by Census (inserts small differences into counts of people and households) to protect each respondent’s identity.
- Invariants (reported as enumerated)
 - Total population (state and state-equivalents level)
 - Total housing units (census block level) – not block-level resident population
 - Number of group quarters by facilities type (census block level) – not population in group quarters



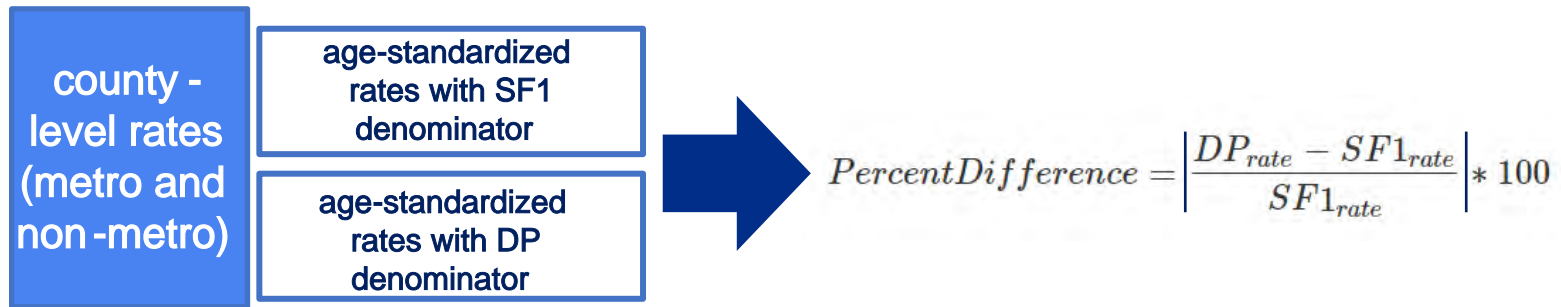
**County level: Assessing the impact
on age-adjusted incidence of COVID-19**

Data and Methods

Data

- Numerator: Number of COVID-19 cases in 2020 by county, race, and age^{*^}
- Denominators: Summary File 1 (SF1) 2010 population data and differential privacy (DP) demonstration data (v4-3-2023)

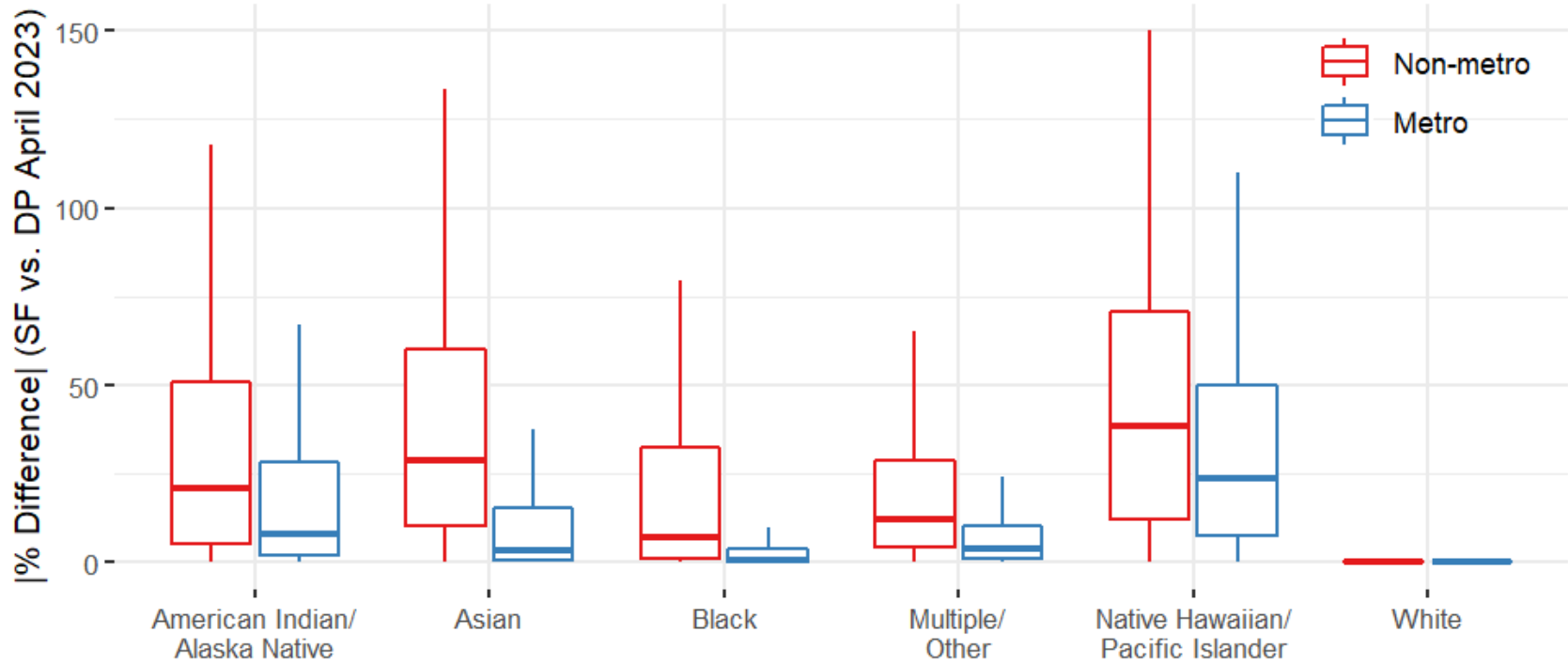
Methods



*included if case report date, age, race, and county of residence were submitted

^limitation for all use cases: 2010 published data used swapping instead of disclosure avoidance

County-level age-adjusted incidence rates for people from racial minority groups were disproportionately affected.

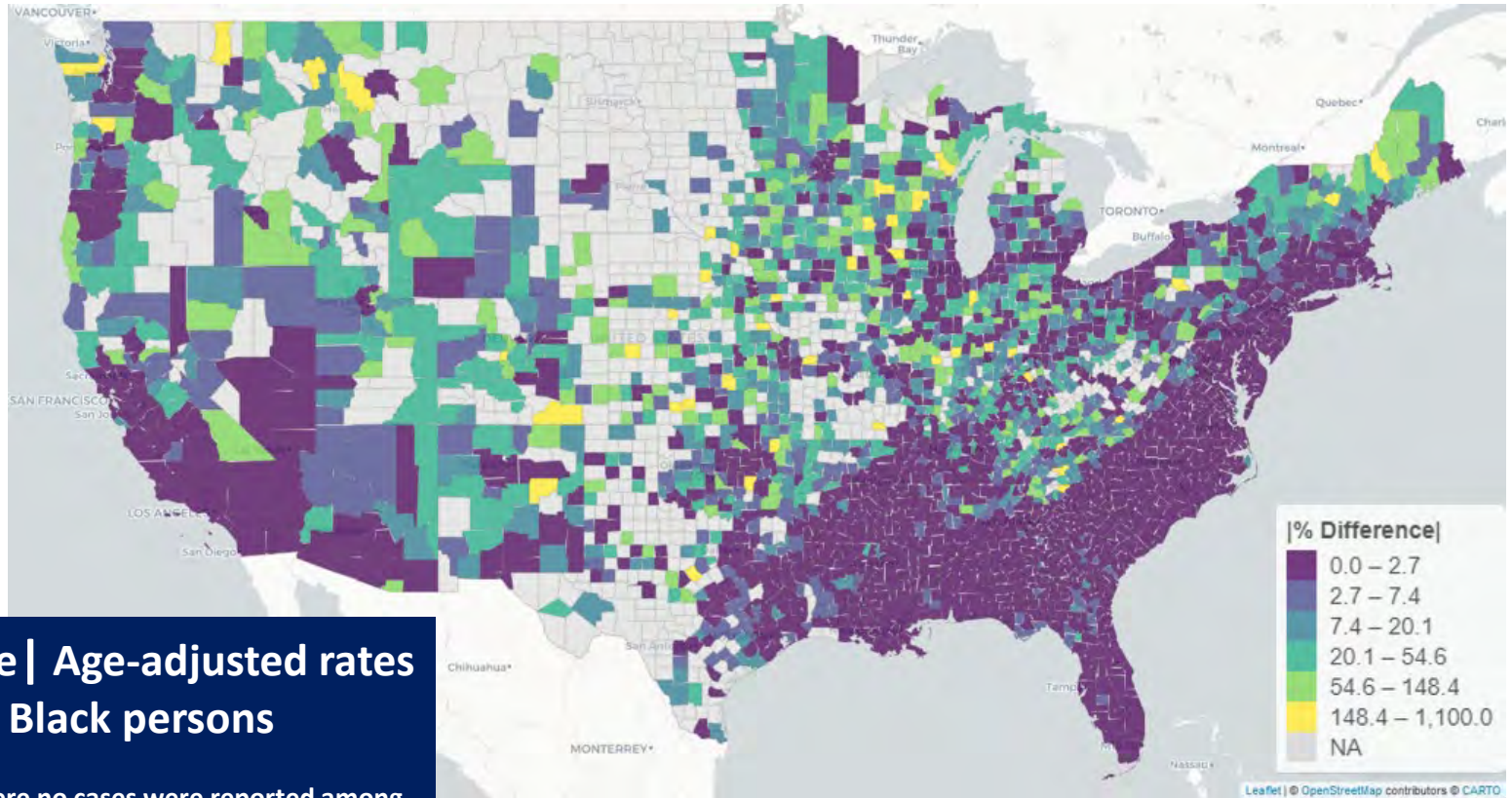


Source: US Census Bureau 2023; CDC Environmental Public Health Tracking Program

Age-adjusted incidence rates for people from racial minority groups were disproportionately affected.

Race/Metro or Non-Metro	Median %	Maximum %
AI/AN – Non-metro	20.8	700
AI/AN – Metro	8.2	700
Asian – Non-metro	28.6	700
Asian – Metro	3.2	559
Black – Non-metro	7.0	900
Black – Metro	0.7	417
Multiple/Other – Non-metro	12.3	1,523.1
Multiple/Other – Metro	3.7	226.1
Native Hawaiian/Other Pacific Islander – Non-metro	38.4	772.9
Native Hawaiian/Other Pacific Islander – Metro	23.8	743.4
White – Non-metro	0.2	52.7
White - Metro	0.04	6.9

COVID-19



**| % difference | Age-adjusted rates
SF vs. DP for Black persons**

NA values occur where no cases were reported among Black persons in that county during 2020.

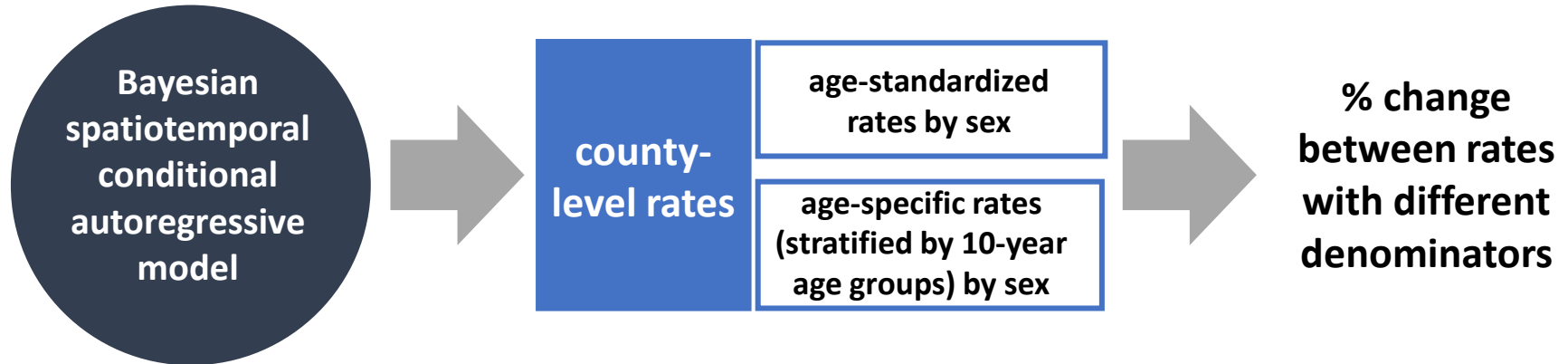
County level: Assessing the impact on heart disease mortality overall and by sub-group

Data and Methods

Data

- Numerator: 2010 county-level heart disease death counts from National Vital Statistics System
- Denominators: National Center for Health Statistics bridged-race estimates and DP demonstration data (v4-3-2023)

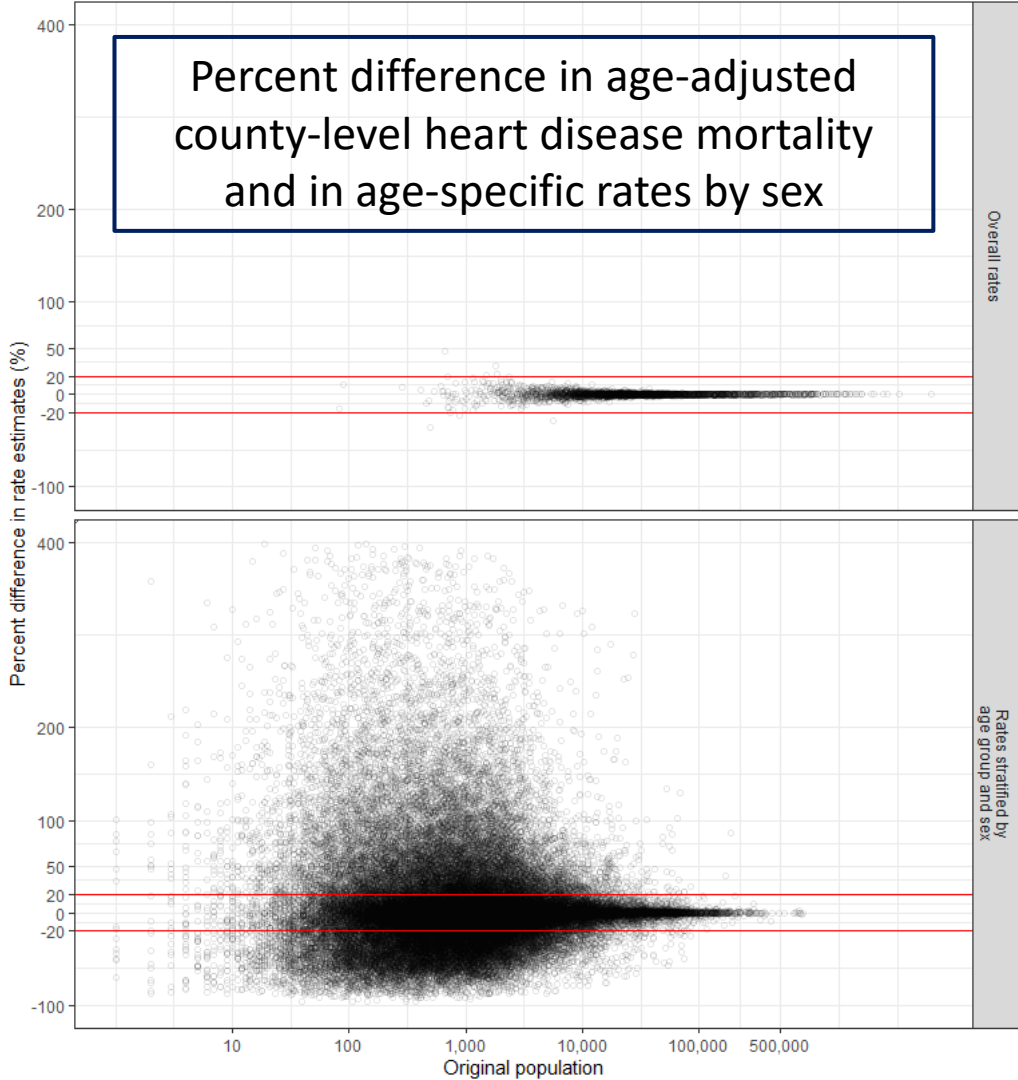
Methods



Heart Disease Mortality

Positive differences indicate the rate estimated using DP was a higher value.

*Outliers truncated.



County and census tract: Assessing the impact on age-adjusted rates of asthma emergency department (ED) visits and acute myocardial infarction (AMI) hospitalizations

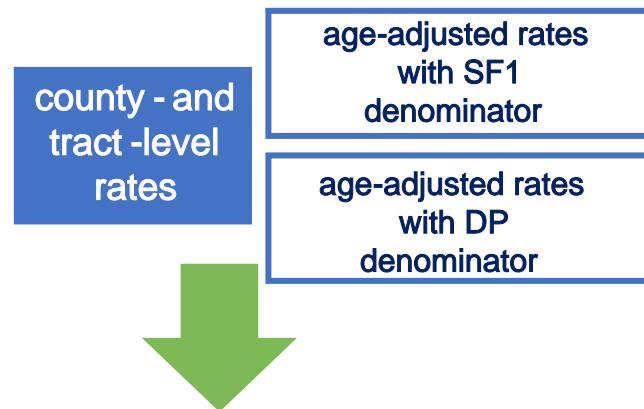
Data and Methods

Data

- **County-level asthma**
 - ED visit counts from 30 states
 - AMI hospitalization counts from 31 states
- **Census tract-level asthma**
 - ED visit counts from 6 states
 - AMI hospitalization counts from 7 states
- **Denominators:** SF1 2010 population data and DP demonstration data (v4-3-2023)

Methods

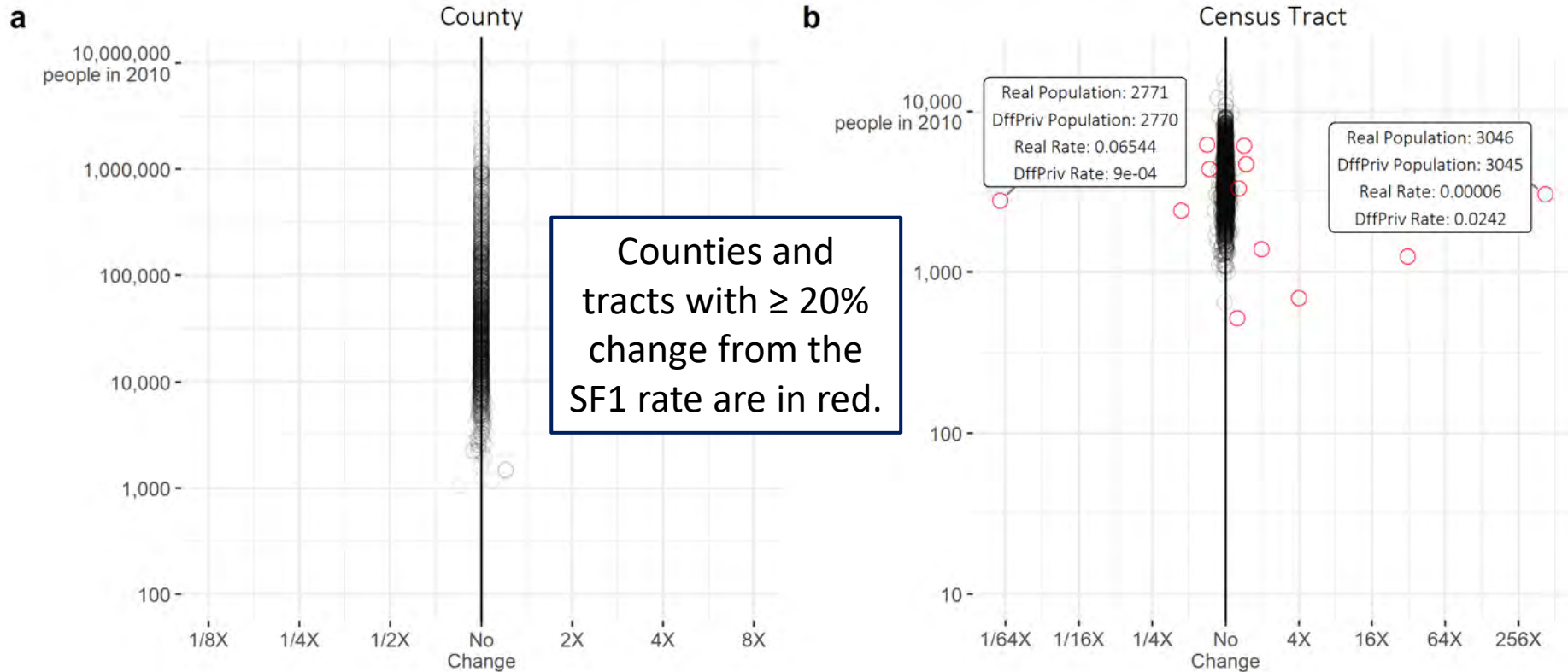
- Calculated age-adjusted rates with and without DP
- Value of percent difference calculated



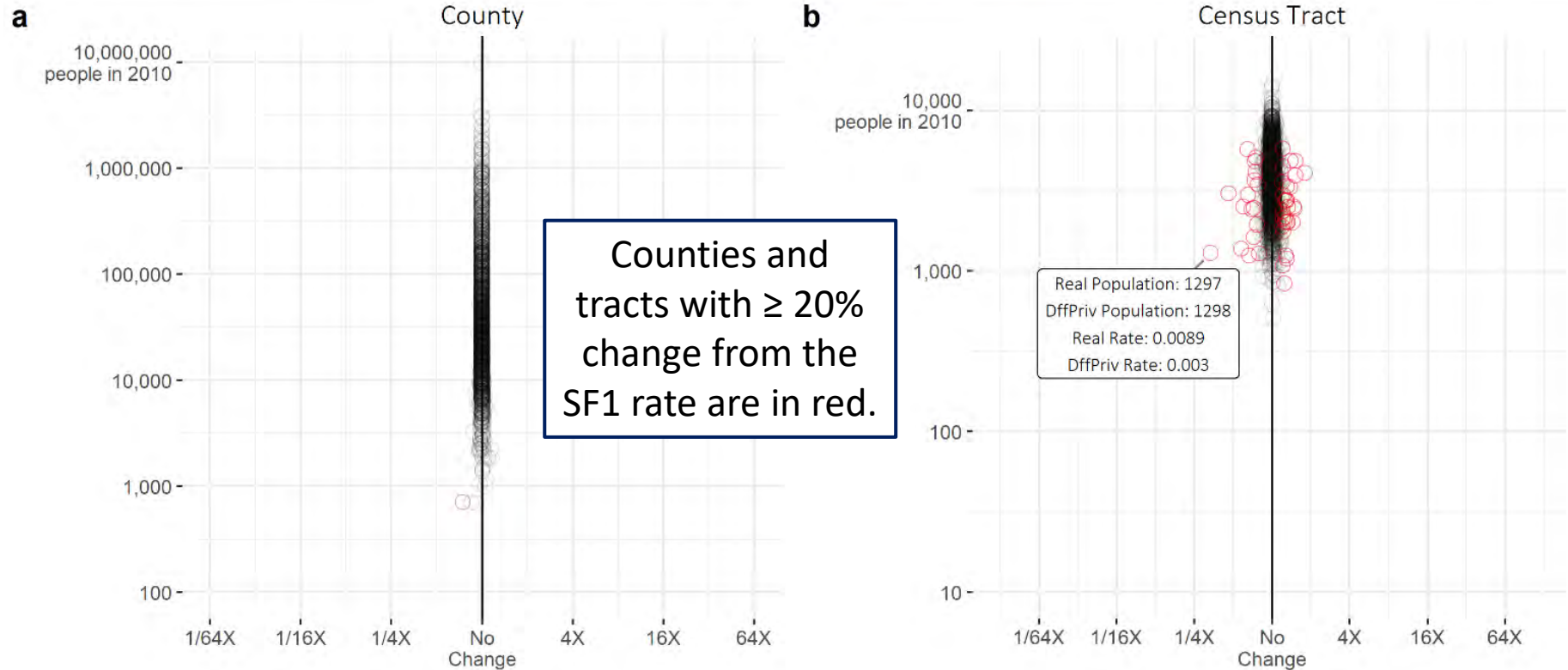
$$\text{PercentDifference} = \frac{DP_{rate} - SF1_{rate}}{SF1_{rate}} * 100$$

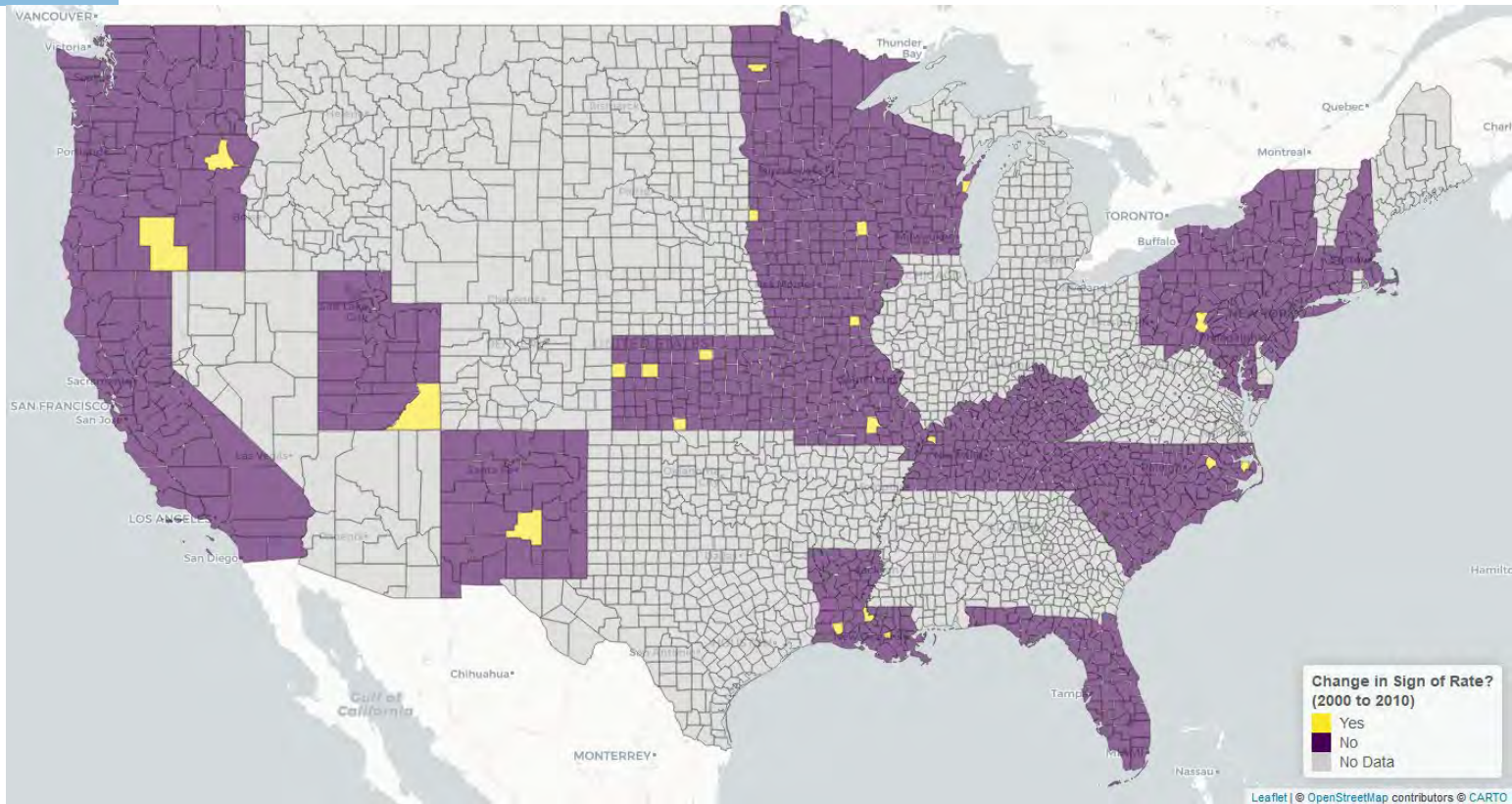
- Evaluated changes in rates between 2000 and 2010 using SF1 and DP denominators

CHANGE IN AGE-ADJUSTED RATE OF ASTHMA ED VISITS (2010)



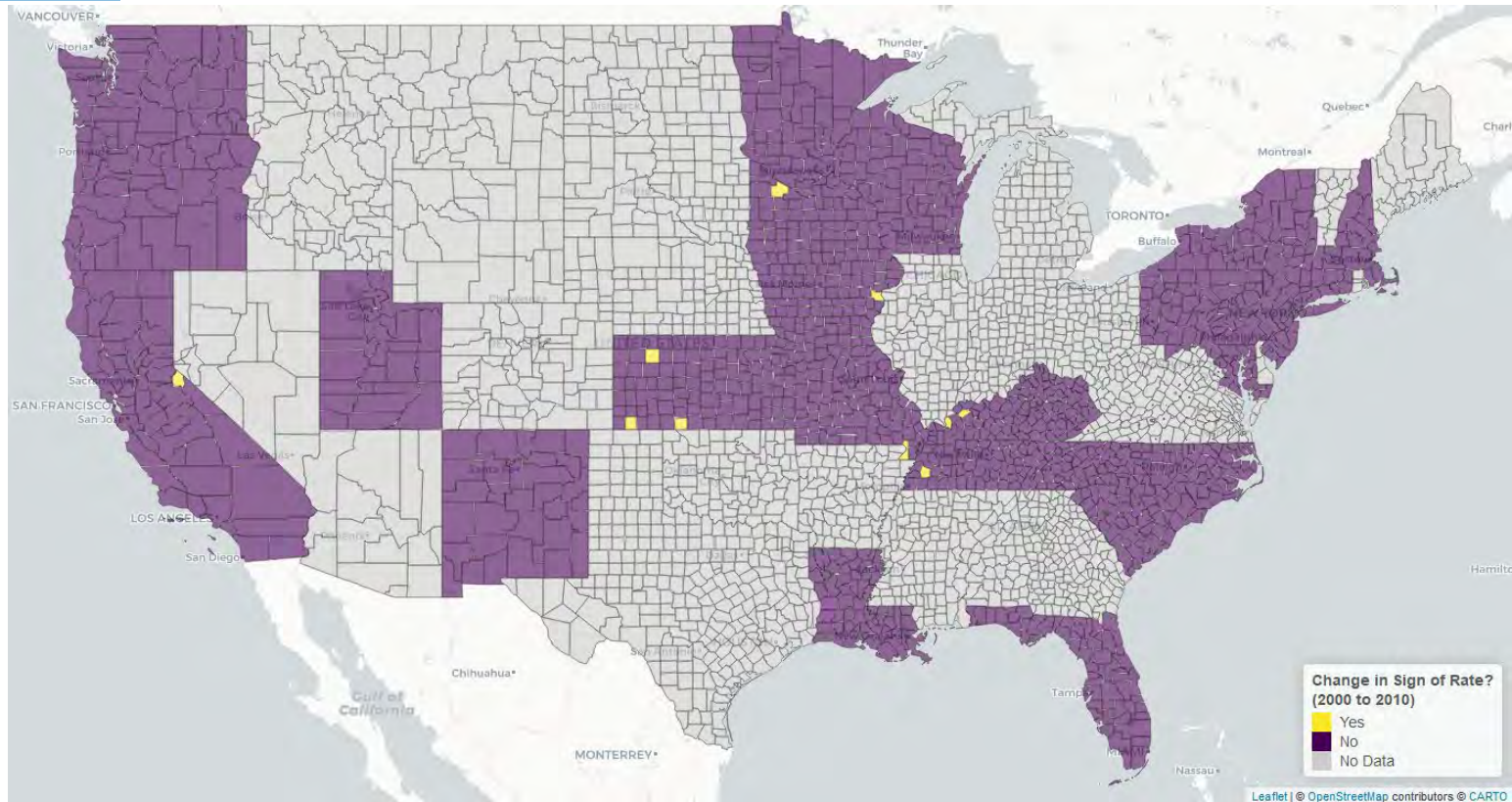
CHANGE IN AGE-ADJUSTED RATE OF AMI HOSPITALIZATIONS (2010)





Counties where the age-adjusted rates of asthma emergency department visits from 2000 to 2010 changed signs due to differential privacy are shown in yellow.

Hosp/ED



Counties where the age-adjusted rates of acute myocardial infarction hospitalizations from 2000 to 2010 changed signs due to differential privacy are shown in yellow.

**Block: Assessing the impact on the Community
Assessment for Public Health Emergency
Response (CASPER)**

Data and Methods

Data

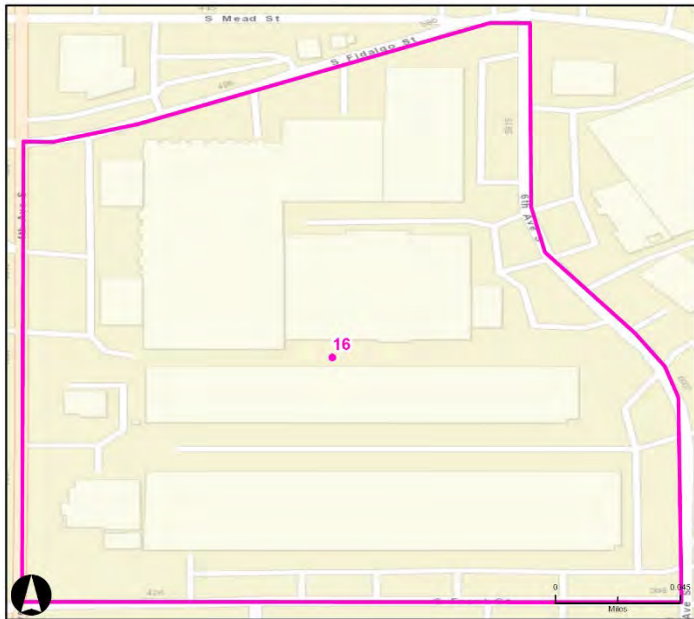
- **2020 Census data at the block level**
 - Occupied households (H3)
- **Uses additional data, where needed**
 - Total population (P1)
 - Population in households by age (P16)
 - Race of householder (H6)
 - Hispanic or Latino origin of householder by race of householder (H7)

Methods

- Block-level data necessary for selecting sample and providing maps to send staff to the field
- Census data used to select 30 “clusters” (blocks) with probability proportional to size
- Seven households interviewed within each cluster
- Survey data are weighted based on 2020 Census data to gain accurate estimates of needs

Community Assessment for Public Health Emergency Response (CASPER)

October 2022



Duwamish

AREA: Georgetown

CLUSTER : 16
INTERVIEWS: 7

Reference Point:
Longitude : -122.32797
Latitude : 47.549321

APPROXIMATE HHs
42

Reference Map



Sample cluster (#16) showing the approximate number of households (42) in the block according to 2020 Census data

Community Assessment for Public Health Emergency Response (CASPER)

October 2022



Duwamish

AREA: Georgetown

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APPROXIMATE HHs
42

Reference Map

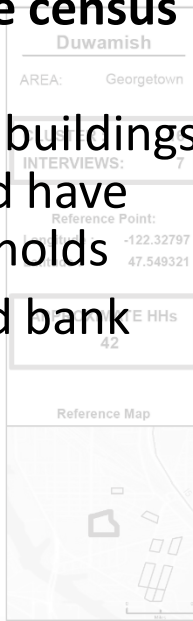


Satellite imagery of the cluster (#16) where 7 interviews should have occurred through systematic selection of the 42 households

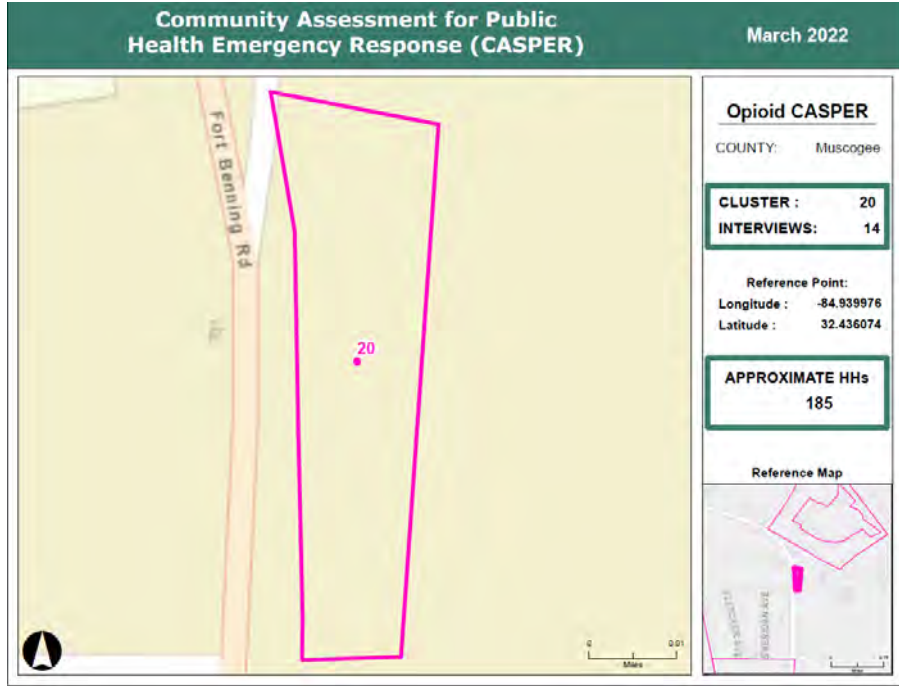
No households actually in the census block

- Assumption that one of the buildings was an apartment and could have accounted for the 42 households
- No households – only a food bank

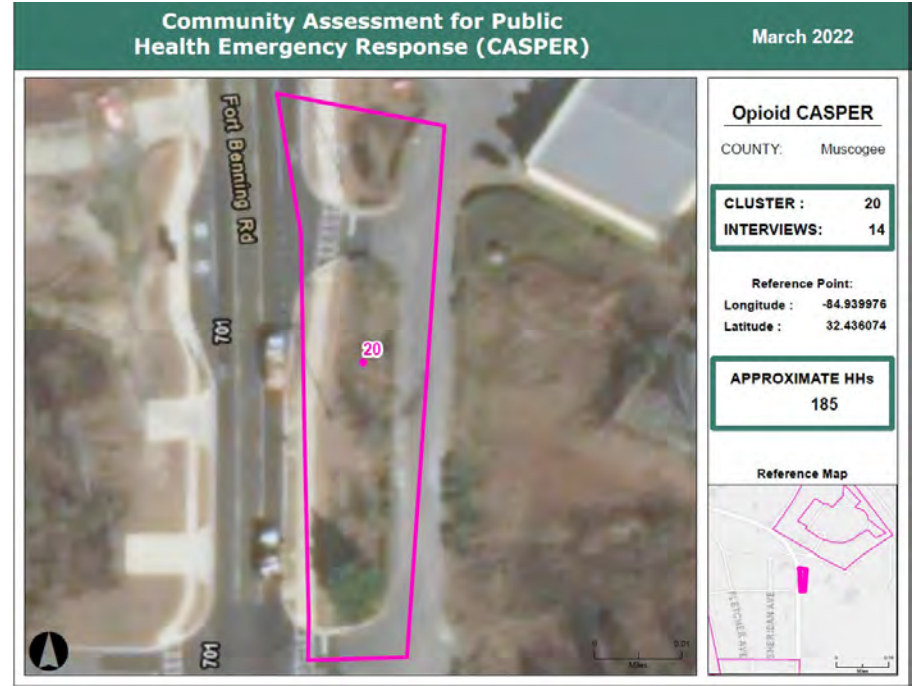
Interviewed a few persons experiencing homelessness



Satellite imagery of the cluster (#16) where 7 interviews should have occurred through systematic selection of the 42 households



Sample cluster (#20) showing the approximate number of households (185) in the block according to 2020 Census data



Satellite imagery of the cluster (#20) where 14 interviews should have occurred through systematic selection of the 185 households

CASPER

- **Teams arrived at cluster expecting 185 households and no households were available to interview**
- **Invalidated the entire assessment**
 - At least one household must be interviewed in each cluster
 - A minimum of 168 households must be interviewed overall
- **CASPER could not be completed after several months of planning and effort**

Community Assessment for Public Health Emergency Response (CASPER) March 2022

CLUSTER : 20
INTERVIEWS: 14

Reference Point:
Longitude : -84.939976
Latitude : 32.436074

APPROXIMATE HHs
185

Reference Map

Community Assessment for Public Health Emergency Response (CASPER) March 2022

Opioid CASPER

COUNTY: Muscogee

CLUSTER : 20
INTERVIEWS: 14

Reference Point:
Longitude : -84.939976
Latitude : 32.436074

APPROXIMATE HHs
185

Reference Map

Satellite imagery of the cluster (#20) where 14 interviews should have occurred through systematic selection of the 185 households

Conclusions From CDC Use Cases

County

- Data showed some overall improvement in age-adjusted rates*
- Significant differences in rates, particularly in counties with smaller populations and when stratifying age-adjusted rates
- Data remain problematic for age-specific rates

**no stratification by sex or by race/ethnicity or urban/rural or any sub-population*

Census tract

- Data remain problematic when calculating age-adjusted rates
- Changes in age groups impact overall rate calculations—even if total population counts don't change significantly

Block

- Data are problematic for characterizing risks accurately and identifying / targeting populations at higher risk
- Can not characterize communities accurately, particularly for situational awareness and emergency response activities

Public Health Implications



Under/overestimation of disease, disparity, or healthcare use rates



**Trend monitoring
over time**

- Detecting true increases or decreases
- Setting targets
- Implementing evidence-based interventions
- Monitoring progress



**Finer spatial resolution
& sub-populations data**

- Disseminating more local-level data
- Targeting populations accurately
- Allocating CDC resources



Measuring baselines and progress of CDC health equity goals

Acknowledgments

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- Amy Schnall, Health Studies Section
- National Center for Health Statistics (NCHS)
- IPUMS

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