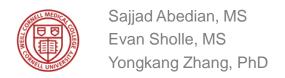
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Augmenting Publicly Available Social Determinants of Health Datasets with Clinical Data to Power Multi-Institutional Clinical Research



Outline

SDH Data Section

- What is SDH?
- How is it collected?
- How to link it?
- Interactive: how to create a SDH analytical file?

Clinical Data Section

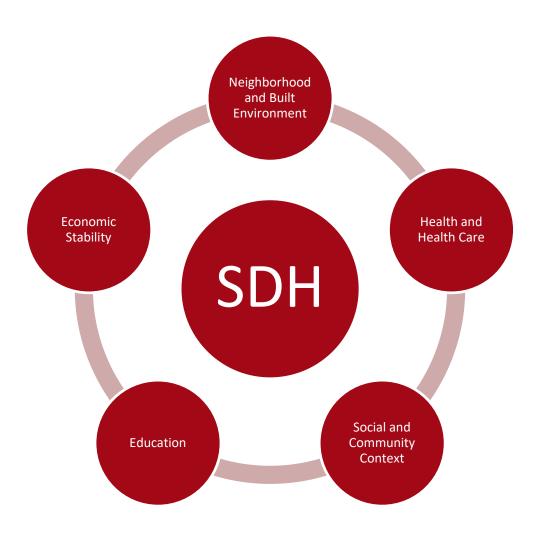
- Interactive: how to create a patient-level analytical file combined with SDH variables?
- Association Between Residential Neighborhood Social Conditions and Health Care Utilization and Costs

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What is SDH?

And why is it important?





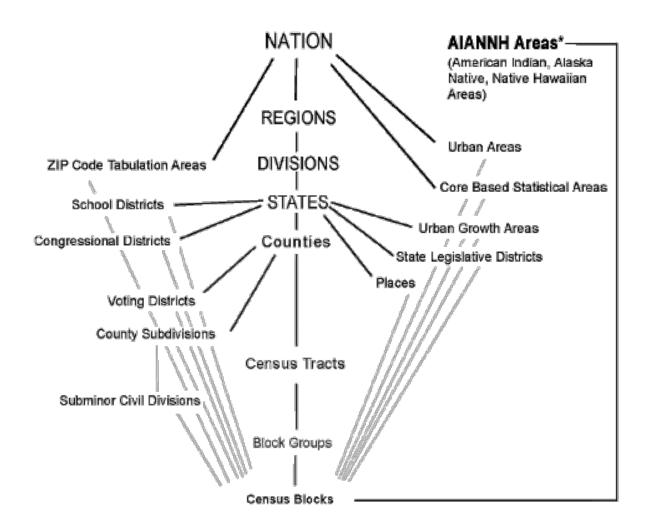
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How is it collected?



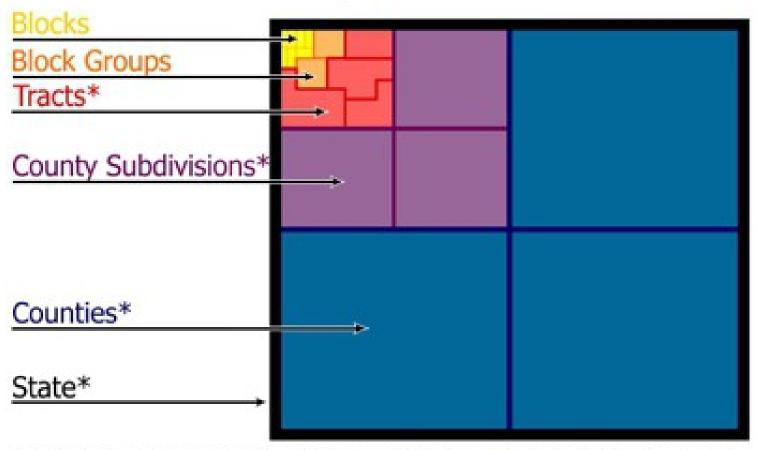
Geographical Units

Standard Hierarchy of Census Geographic Entities



Census Tract vs. Block Group

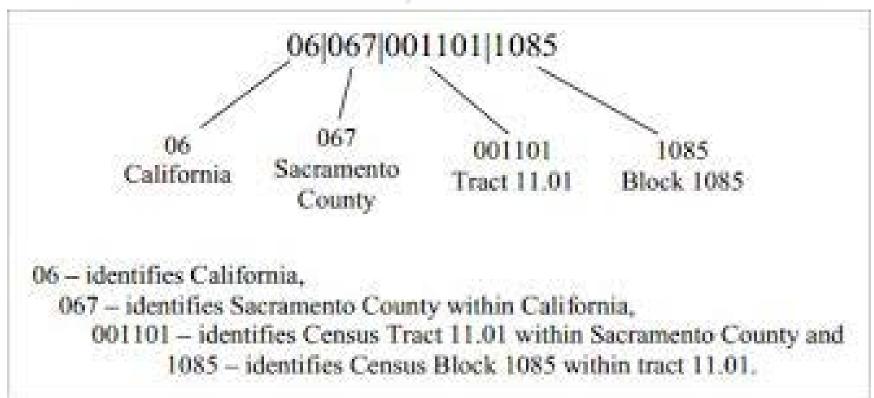
Census Summary Levels:



^{*}Included with Maptitude. Blocks and block groups available separately.

Census Tract vs. Block Group

Census Summary Levels:

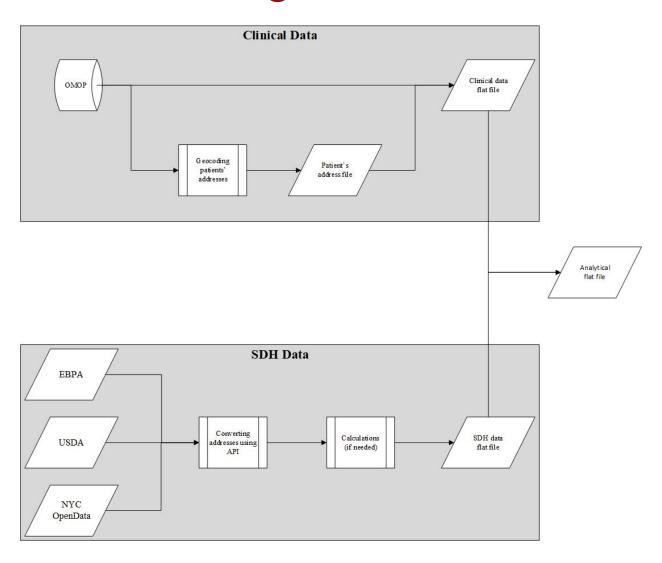


^{*}Included with Maptitude. Blocks and block groups available separately.

Sources

- Publicly available data sources
 - American Community Survey (ACS)
 - EPA National Air Toxics Assessment
 - USDA
 - NYC OpenData
 - Factors Affecting Communities and Enabling Targeted Services (FACETS)
- EHR
 - Structured
 - Unstructured

Process of Linking EHR to SDH





VACCINe

Variables Affecting Care in Communities and Individuals in New York

https://github.com/SajjadAbedian/AMIA 2020 VACCINE/blob/main/VACCINe walkthrough.ipynb





Association Between Residential Neighborhood Social Conditions and Health Care Utilization and Costs

Yongkang Zhang, PhD

11/14/2020



Disclosure

I disclose that neither I nor my partner have relevant financial relationships with commercial interests.

Learning Objectives

- Understand strategies to combine social determinants of health data, clinical data, and administrative claims data for health services research.
- Understand the associations between disadvantaged neighborhood social conditions and healthcare utilization and costs among Medicare patients.

SDoH and Healthcare

- SDoH have strong associations with healthcare quality, utilization, and patient outcomes
 - Social factors account for 25%–60% of deaths in the United States
 - Neighborhood social disadvantage is associated with increased risk of 30-day readmission
 - Food-insecure children are twice as likely to report being in fair or poor health and at least 1.4 times more likely to have asthma, compared to food-secure children

Efforts to Address Social Risk Factors

- A growing interest in addressing social risk factors to improve population health among policy makers, researchers, and healthcare providers
 - Comprehensive screening to identify unmet health-related social needs
 - Increase patients' awareness of community services
 - Navigation services to assist high-risk patients with accessing community services
 - Alignment between clinical and community services to ensure that community services are available and responsive to the needs of patients

Gaps in Literature

- Many studies have found that disadvantaged social conditions are associated with poor clinical outcomes and hospital readmissions
- Little is known about how SDoH are associated with other types of potentially preventable healthcare utilization and costs associated with preventable utilization
- Limited evidence is available regarding how SDoH influence total healthcare utilization or costs

Objectives of the Study

 To examine the associations of neighborhood social conditions with total annual and potentially preventable Medicare costs

Methods: Population and Sample

- Medicare fee-for-service patients who had at least one clinical encounter in five major health systems in New York City
 - Patients with a valid 9-digit zip code in New York or New Jersey
 - Continuously enrolled in Medicare Parts A and B for 12month in 2013 and 2014
 - Had no Medicare Advantage enrollment
 - 93,429 patients were included

Methods: Healthcare Utilization and Costs

Medicare claims data

- Healthcare utilization across all care settings and different providers and organizations
- Payment made by Medicare, patients, and other payers
- Importance of using claims data to measure utilization and costs (Smith et al., JAMIA, 2019).

Total Medicare costs

 Standardized total annual Medicare costs across all service categories for 2013 and 2014, including physician services, outpatient services, inpatient services, durable medical equipment, Part D prescription, and post-acute care (e.g., skilled nursing facility and home health) and hospice services

Methods: Healthcare Utilization and Costs

- Potentially preventable utilization
 - Preventable ED visits
 - —Billings' algorithm
 - Preventable hospitalizations
 - —AHRQ's PQI indicators
 - Unplanned 30-day hospital readmissions
 - —CMS algorithm for pay-for-performance programs
- Potentially preventable costs
 - Standardized episodic costs of all potentially preventable utilization and all services delivered within 30 days after a preventable ED visit or hospitalization for each patient in 2013 and 2014.

Methods: Demographics, Diagnoses, and Procedures

- Medicare claims data across all care settings
- Electronic health records data for the same patient cohort from multiple health systems in NYC.

- Multiple choices of SDoH from different resources
 - o Individual or community level?
 - —Medicare claims and EHR data have very limited individual level SDoH: dual-eligible status
 - Comprehensive SDoH are available at community level
 - o What level of community?
 - Medicare claims data contain state, county, and 5-digit zip codes of patient residential location
 - -9-digit zip codes are available from EHR data



- Multiple choices of SDoH from different resources
 - Data available with different granularity

Neighborhood	Distinct Areas
State	50
County	3,200
Zip Code Tabulation Area	32,000
5-digit Zip Code	42,000
Census Track	74,000
Census Block Group	221,000
Census Block	11 millions
Possible 9-digit Zip Code	30 millions



- Many policy and economic studies used state for country level SDoH data
- Heath policies are developed and implemented at the state or county level
- Policy makers need to understand how state or county level SDoH are associated with uptake of health policies among healthcare organizations and providers
 - Health information technology
 - Novel payment and delivery models (e.g., ACO)
- State or country level SDoH data may be important confounders that need to be adjusted in policy analysis



	ZIP (Zone Improvement Plan) code	ZCTA (ZIP Code Tabulation Area) code
Developer	United States Postal Service (USPS)	United States Census Bureau
Year of Implementation	5-digit: 1963 9-digit zip+4: 1983	Census 2000
Total Numbers	42,000	32,000
Purpose	Introduced as as a method to improve mail delivery service	Tabulate summary statistics
Data availability	Limited	5-year American Community Survey
Boundary	Could cross states	In a few cases, ZCTAs can cross into bordering states
Frequency of change	ZIP codes change frequently. Every year there are thousands of changes to the USPS ZIP Code records.	Could be stable for 10 years or more.



- Census block groups: small, relatively homogenous areas of 600-3,000 persons
- Census tracts: aggregates several block groups, and represent approximately 4,000 persons
- Many SDoH measures are available at both tract and block group level
 - American Community Survey

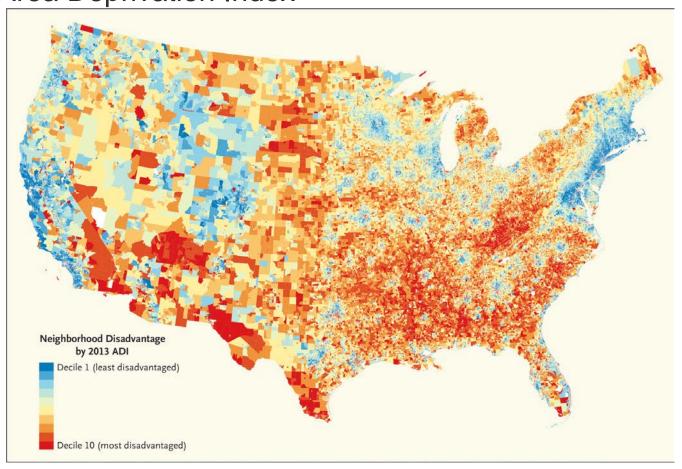
- Multiple choices of SDoH from different resources
 - Composite score or individual measures?
 - —Individual measures reflecting important neighborhood socioeconomic status: median household come, unemployment rate, % of people without high school degree etc.
 - Composite scores: a summary of neighborhood social conditions across different measures: Area Deprivation Index, Social Deprivation Index, Social Vulnerability Index etc.

- Multiple choices of SDoH from different resources
 - Area Deprivation Index

·	
Components for Area Deprivation Index	Weight
Percentage of population aged ≥ 25 with < 9 years of education	0.0849
Percentage of population aged ≥ 25 with at least a high school diploma	-0.0970
Percentage of employed persons aged ≥ 16 in white collar occupations	-0.0874
Median family income	-0.0977
Income disparity (log of 100 * ratio of the number of households with $<$ \$10,000 annual income to the number of households with \ge \$50,000 annual income)	0.0936
Median home value	-0.0688
Median gross rent	-0.0781
Median monthly mortgage	-0.0770
Percentage of owner-occupied housing units (home ownership rate)	-0.0615
Percentage of civilian labor force population aged ≥ 16 unemployed (unemployment rate)	0.0806
Percentage of families below the poverty level	0.0977
Percentage of population below 150% of the poverty threshold	0.1037
Percentage of single-parent households with children aged < 18	0.0719
Percentage of occupied housing units without a motor vehicle	0.0694
Percentage of occupied housing units without a telephone	0.0877
Percentage of occupied housing units without complete plumbing	0.0510
Percentage of occupied housing units with >1 person per room	0.0556

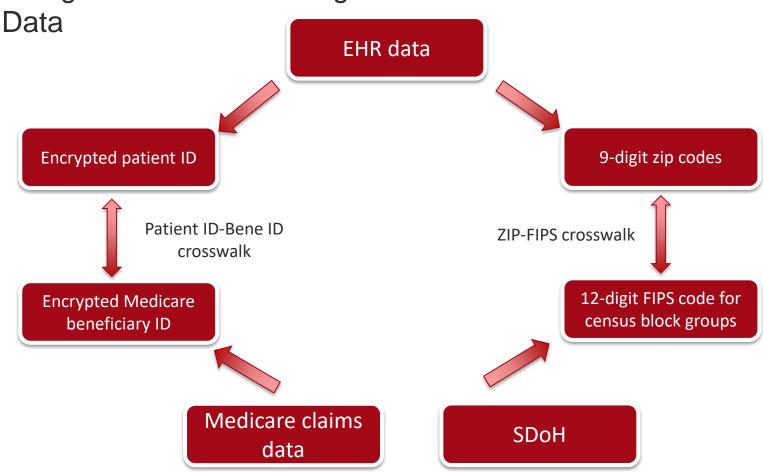
Multiple choices of SDoH from different resources

Area Deprivation Index



Methods: Data Linkage

Using EHR data as a bridge between SDoH and Medicare



Analysis: Descriptive Analysis

- We first compared differences in demographics, comorbidities, and other patient characteristics across patients in different ADI quintiles
 - Demographics: age, sex, and race
 - Comorbidities: number of chronic conditions, frailty, serious mental illness, serious physical illness, and endstage renal disease
 - CMS Hierarchical Condition Categories score for each patient to measure the.
 - Dual-eligible status and Medicare Part D coverage
- We used χ2 tests for categorical variables and t-tests for continuous variables to test for differences in patient characteristics across ADI quintiles.

Analysis: Multivariable Analysis

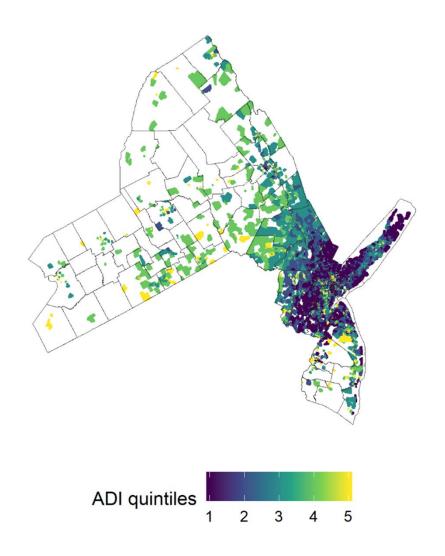
- Generalized linear models (GLM) to examine the association of ADI quintiles with (1) Total annual Medicare costs and (2) Cost associated with preventable utilization
- We also tested the association between ADI quintiles and the probability of having the three types of the preventable utilization by using logistic regressions

Analysis: Multivariable Analysis

- We focused on the neighborhoods with the least disadvantaged conditions (quintile 1) and the most disadvantaged conditions (quintile 5), as compared with the neighborhoods with the intermediate level conditions (quintile 3)
- All regressions were adjusted for demographic, comorbidities, and other patient characteristics of each year
- We additionally added year-fixed effects to control for secular changes in outcomes and indicators for hospital referral regions to account for supply side variations across regions

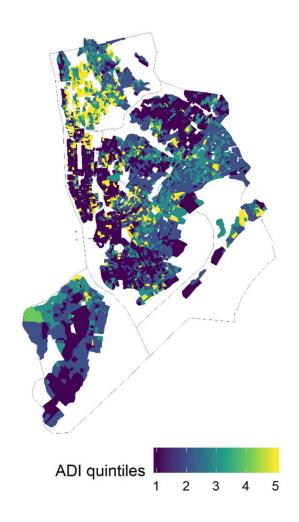
Results: Descriptive Analysis

Geographic location and social conditions



Results: Descriptive Analysis

Geographic location and social conditions



Results: Descriptive Analysis

 Patients with different social conditions have different characteristics.

		Quintiles of ADI at Census Block Group Level				
Patient and Neighborhood Characteristics	All (N = 93,429)	Quintile 1 (N = 48,974)	Quintile 2 (N = 17,838)	Quintile 3 (N = 13,480)	Quintile 4 (N = 5010)	Quintile 5 (N = 8127)
Neighborhood characteristics						
No. census block groups included in this study	12,420	3849	3516	2690	1102	1263
Average number of Medicare or dual-eligible residents in each block group*	160.0	156.0	159.4	163.0	167.3	152.1
Median family income (\$)*	82,197	133,891	91,750	77.917	66,008	41,197
Patient characteristics	02,127	100,001	71,700	,,,,,,,	00,000	12,227
Age (%)						
< 65	12.4	7.2	16.0	18.0	16.9	24.1
65–74	40.1	42.2	40.8	37.5	37.2	32.0
75–84	32.3	34.5	29.9	29.2	30.1	29.5
≥85	15.2	15.9	13.4	15.3	15.8	14.4
Sex (%)						
Male	39.4	40.1	40.6	38.8	37.5	35.1
Female	60.6	60.0	59.4	61.2	62.5	64.9
Race (%)						
White	79.6	88.3	76.6	73.2	73.1	48.9
Black	9.7	4.1	12.0	14.6	14.9	27.5
Hispanic	3.2	1.2	3.2	4.5	4.8	11.7
Other	7.5	6.5	8.3	7.7	7.2	11.9
Dual eligible (%)	23.6	13.5	25.7	32.5	27.9	61.8
Part D coverage (%)	76.8	74.4	77.1	78.7	77.5	87.1
ESRD	1.9	1.0	2.6	2.8	2.0	4.6
Originally enrolled in Medicare because of disability	19.9	12.8	24.9	27.3	25.5	35.8
No. chronic conditions (%)						
0	3.5	3.4	3.3	3.2	4.0	4.1
1–2	10.6	11.6	9.2	9.5	10.8	9.3
3–5	33.3	36.1	30.8	29.8	32.5	28.9
6–8	32.7	31.9	34.2	33.8	31.9	32.4
≥9	20.0	17.0	22.5	23.8	20.9	25.3
Frailty (%)	18.6	16.9	19.4	20.9	20.0	22.9
Mental illness (%)	21.2	18.8	20.8	24.7	23.9	28.7
Seriously ill (%)	17.1	15.2	18.7	19.1	18.5	20.5
HCC score, mean	1.5	1.3	1.6	1.6	1.5	1.7

We compared the differences in patient characteristics between quintile 1 and each other quintile (quintiles 2-5) using t test and the difference are all statistically significant (P < 0.001).

^{*}Data are from American Community Survey to represent the overall characteristics of these communities. ADI indicates Area Deprivation Index; ESRD, end-stage renal disease; HCC, Hierarchical Condition Categories.

Results: Descriptive Analysis

 Patients with disadvantaged social conditions have higher unadjusted Medicare spending

	Average total annual Medicare costs (\$)	Physician (\$)	Outpatient (\$)	Inpatient (\$)	Post-acute care and hospice (\$)	DME (\$)	Part D (\$)
ADI, quintile 1	17,661	5,928	2,460	3,744	1,732	230	3,566
ADI, quintile 2	22,965	6,880	3,210	5,260	2,400	325	4,890
ADI, quintile 3	23,841	6,867	3,311	5,305	2,665	336	5,356
ADI, quintile 4	21,564	6,174	2,965	4,830	2,567	312	4,716
ADI, quintile 5	26,815	6,174	4,154	6,001	2,969	432	6,860
Average	20,670	6,398	2,901	4,514	2,147	285	4,425

Results: Descriptive Analysis

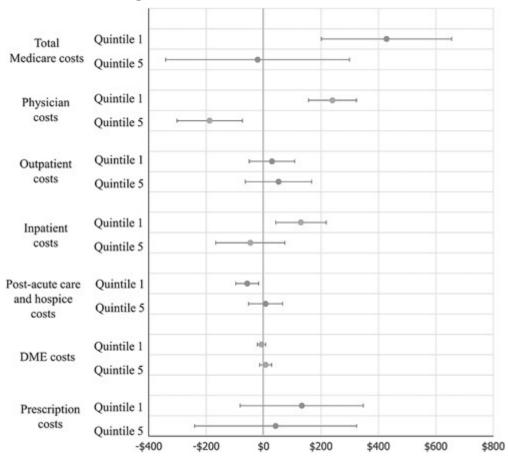
 Patients with disadvantaged social conditions have higher unadjusted preventable spending and utilization

	Potentially preventable Medicare costs (\$)	Preventable ED visits (%)	Preventable hospitalizations (%)	30-day readmissions (%) *
ADI, quintile 1	1,343	8.7	3.7	13.7
ADI, quintile 2	2,260	11.9	5.4	16.8
ADI, quintile 3	2,568	13.3	6.1	18.3
ADI, quintile 4	2,232	13.5	5.6	18.9
ADI, quintile 5	3,361	18.5	8.1	20.0
Average	1,918	11.1	4.8	16.3

^{*} Analysis only using hospitalized patients

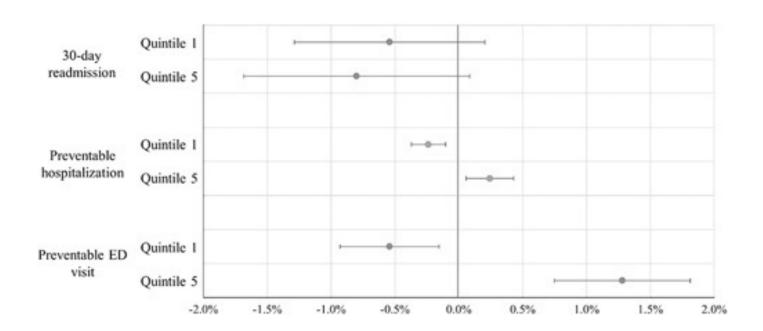
- Compared with the neighborhoods with intermediate level social conditions (quintile 3), the least disadvantaged neighborhoods (quintile 1) was associated with higher total Medicare costs (\$427 or 3.2% higher, P<0.001)
- The most disadvantaged neighborhoods (quintile 5) had similar total Medicare costs (\$22 or 0.2% lower, P=0.89) after adjusting for patient demographics, comorbidities, and other characteristics

The association between social disadvantage and spending varies by care setting



- Compared with the neighborhoods with intermediate level social conditions (quintile 3), the least disadvantaged neighborhoods (quintile 1) had similar potentially preventable costs (-\$23; 95% CI, -\$56 to \$10)
- The most disadvantaged neighborhoods (quintile 5) had higher potentially preventable costs (\$53; 95% CI, \$1-\$104) than patients with the intermediate level social conditions

The association between social disadvantage and preventable healthcare utilization



Discussions

- Disadvantaged neighborhood conditions are associated with lower total annual Medicare costs but higher potentially preventable costs after controlling for demographic, medical, and other patient characteristics.
- Socioeconomic barriers may limit access and use of primary care and disease management services, resulting in a higher proportion of their health care costs going to potentially preventable care.

Discussions

- Results of this study do not necessarily indicate causal relationships between social conditions and health care utilization and costs
- This study was not able to incorporate individual level SDoH in the analysis and therefore cannot examine if the associations of community SDoH are still significant after controlling for individual level SDoH
- The 3 types of preventable utilization may not all be truly preventable; alternatively these services may not represent the totality of preventable utilization

Acknowledgement

- Patient-Centered Outcomes Research Institute Health Systems PCORnet (Patient-Centered Clinical Research Network) Demonstration Project (HSD-1604-35187, PI: Kaushal)
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Thank you!

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