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Healthcare Hotspotting – A Randomized Controlled Trial

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Abstract

Background: There is widespread interest in programs aiming to reduce spending and improve quality among “super-utilizers,” patients with very high use of healthcare services. The Camden Coalition of Healthcare Providers’ (the Coalition) “Hotspotting” program has received national attention as a promising super-utilizer intervention and has been expanded to cities around the country. In the months following hospital discharge, a team of nurses, social workers and community health workers visit enrolled patients to coordinate outpatient care and to link them with social services.

Methods: We randomized 800 medically and socially complex hospitalized patients, all with at least one additional hospitalization in the prior six months, to the Coalition’s care transition program or to usual care. The primary outcome was hospital readmission within 180 days post-discharge.

Results: The 180-day readmission rate was 61.7 percent in the control group and 62.3 percent in the intervention group. The adjusted difference between the groups was not significant (0.82 percentage points, 95% CI –5.97 to 7.61). By contrast, comparing the intervention-group admissions during the six months before and after enrollment misleadingly suggested a 38 percentage point decline in admissions from the intervention, because it did not account for the similar decline in the control group.

Conclusions: In this randomized controlled study of patients with very high use of healthcare services, readmission rates were not lower for patients randomized to the Coalition’s program compared to usual care. (Funded by the National Institute on Aging, J-PAL North America, and MIT Sloan School of Management; [ClinicalTrials.gov](https://clinicaltrials.gov) number, and the American Economic Association registry number, AEARCTR-0000329.)

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Introduction

Healthcare spending in the United States is heavily concentrated. Five percent of the population accounts for 50 percent of annual spending; one percent accounts for almost one-quarter of annual spending.¹ There is therefore substantial interest in interventions that can reduce spending and improve quality by targeting “super-utilizers” of the healthcare system. Such programs have received considerable positive media attention^{2–7} as well as support from the federal government.^{8,9}

Since being profiled in Atul Gawande’s seminal *New Yorker* article, “The Hot Spotters”,¹⁰ the Camden Coalition of Healthcare Provider’s (the Coalition’s) program has been the flagship example of a promising super-utilizer program. The Coalition’s Camden Core Model uses real-time data on hospital admissions to identify super-utilizer patients, an approach referred to as “hotspotting.” Focusing on patients with chronic conditions and complex needs, and starting with the premise that the standard system is difficult to navigate for these patients, the program uses a high-touch, face-to-face care model to engage patients and connect them to appropriate medical care, government benefits, and community services, with the aim of improving health and reducing unnecessary utilization.

The program has been heralded as a promising, data-driven, relationship-based, intensive care management program for super-utilizers, and federal funding has expanded versions of the model to other cities.^{7–16} To date, however, the only evidence of its impact is a pre-post analysis of the healthcare spending of 36 participants¹⁷ and an evaluation of four expansion sites comparing 149 program participants with propensity-score matched controls.¹⁸ More broadly, there are a number of promising observational studies of other super-utilizer programs.^{12,17,19–21} However, regression to the mean—the tendency for patients selected as exceptionally high-cost at a moment in time to move closer to average cost over time—may bias observational studies of super-utilizer programs towards spurious results.^{22,23}

Although there is limited rigorous evidence of the effectiveness of super-utilizer programs, several randomized trials of care transition programs—which, like the Camden Core Model, start with patients in the hospital and work with them post-discharge—find substantially reduced readmissions.^{24–29} However, the Camden Core Model targets a much more heterogeneous population with greater social and medical complexity and substantially higher healthcare utilization. Therefore, the Coalition partnered with the investigators to design a prospective randomized evaluation of this nationally-recognized program.

Methods

Study Design

This trial was an investigator-initiated, randomized controlled study. It received IRB approval from Cooper University Hospital, the National Bureau of Economic Research, Kennedy Health, and Our Lady of Lourdes Medical Center; it was registered at clinicaltrials.gov³⁰ () and the American Economic Association registry (AEARCTR-0000329).³¹ The trial protocol and planned analyses were publicly pre-specified in March 2014³¹ in consultation with Dr. Brenner, then director of the Coalition.

Minor departures from the pre-analysis plan are described in the supplemental appendix. The Coalition staff implemented the protocols and administered the intervention for the treatment group, but was blinded to results prior to trial completion.

Program

Eligibility: The Camden Core Model is a care transition program designed to improve patient health and reduce hospital use among some of the least healthy and most vulnerable adults in the United States. Eligibility is limited to adults (ages 18 to 80) living in Camden, NJ, one of the most economically depressed and violent cities in the country;¹⁰ in 2017, 37 percent of Camden residents lived below the poverty line, compared to 15 percent nationally.³²

The intervention targeted super-utilizers of the healthcare system—individuals with medically and socially complex needs who have frequent hospital admissions. The inclusion criteria were: at least one hospital admission at any of four Camden-area hospital systems in the six months prior to the index admission when patients were enrolled, at least two chronic conditions, and at least two of the following: at least five active outpatient medications, difficulty accessing services, lack of social support, a mental health comorbidity, an active drug habit, or homelessness. Patients were excluded if they were uninsured, cognitively impaired, an oncology patient, or admitted for a surgical procedure for an acute problem, for mental health care (with no comorbid physical health conditions), or for complications of a progressive chronic disease with limited treatments. The eligible population was less than one-half of one percent of the Camden population but accounted for 11 percent of Camden hospital expenditures (see supplementary appendix).

Intervention: The time-limited intervention had intensive clinical and social components. Patients were enrolled while in the hospital. Once they returned home, they were engaged by a multidisciplinary team including registered nurses, social workers, licensed practical nurses, community health workers, and health coaches. The team conducted home visits, scheduled and accompanied participants to initial primary and specialty care visits, coordinated follow-up care and medication management, conducted blood pressure and blood sugar checks, coached participants in disease-specific self-care, and helped participants apply for social services and appropriate behavioral health programs. The intervention contained many characteristics considered important for successful care transition programs for high-cost, high-need patients.^{33,34} See supplemental appendix for more detail on the intervention.

The control group received usual post-discharge care, that may have included home healthcare services or other outreach; we are unable to measure such services.

Recruitment and Randomization protocols

The study added consent and randomization to the Coalition's pre-existing protocol. Recruitment took place at the two major hospitals: Cooper University Hospital and Our Lady of Lourdes Hospital. Using the Camden Coalition Health Information Exchange database—which provided daily updates from hospital electronic medical records from

Cooper, Lourdes, Virtua, and Kennedy Health (as of July 2014)—staff selected potentially eligible patients, who formed the triaged population. A Coalition recruiter approached triaged patients in the hospital, confirmed eligibility, obtained informed consent, and conducted a baseline survey. The recruiter then used a tamper-proof and externally-recorded randomization process to assign treatment or control status and informed the patient. All patients completing the baseline survey were compensated \$20 for their time. See supplemental appendix for more details.

We piloted the study from March 29, 2014 to May 30, 2014. The study population was enrolled from June 2, 2014 through September 13, 2017.

Of the 1,520 patients triaged, recruiters deemed 1,442 eligible and consented 809; half were randomized to treatment. Subsequently, five consented patients were excluded at their request; the last four patients enrolled were excluded to reach the target study population of 800 (Figure 1).

Data

The primary data were hospital discharge data through March 31, 2018 from the four Camden hospital systems; these accounted for 98 percent of NJ hospital discharges for Camden residents (see supplemental appendix). The discharge data contained admission and discharge dates, diagnoses, discharge destination, charges and payments received, as well as patients' identifying information.

We supplemented these data with several sources. The Camden Coalition Health Information Exchange database contained additional demographic information and a record of the study participant's index admission (where recruitment occurred). We matched 782 (98 percent) to the discharge record for their index admission; match rates were balanced between treatment and control (Table 1). The baseline survey provided additional socio-economic information on consented patients. The Coalition recorded staff engagements with patients in the treatment group. NJ administrative data measured social services receipt (specifically Supplemental Nutrition Assistance Program, Temporary Assistance for Needy Families, and General Assistance) and the National Death Index provided mortality data. See supplemental appendix for additional details.

Outcomes

The primary outcome was readmission within 180 days post discharge. Secondary outcomes were the number of readmissions, proportion of patients with 2 or more readmissions, hospital days, charges, payments received, and mortality—all measured 180 days post discharge—as well as readmission rates at other time horizons. We also analyzed the primary outcome by pre-specified subgroups. With the exception of social services receipt and mortality, all outcomes were based on hospital discharge data.

Statistical Analyses

We used linear regressions to compare outcomes for treatment and control patients. To increase precision, we included pre-specified covariates: age (in five-year bins), gender,

indicators for non-Hispanic African American and for Hispanic, and measures of utilization in the 0–6 and 7–12 months prior to the index admission. We also report the unadjusted difference. We conducted sensitivity analysis using multiple imputation to account for missing outcome data for 18 participants who could not be matched to the discharge record for their index admission.³⁵

Prior to the pilot, we calculated that a study population of 800 would provide statistical power to detect a 9 percentage point decline in the 180 day readmission rate (80% power, two-sided test size 0.05).³¹ Data from the pilot on the actual study population—whose readmission rate was twice what we had assumed—indicated power to detect a 9.6 percentage point decline in the primary outcome (see supplemental appendix). There was no pre-specified plan to adjust for multiple comparisons; therefore, we report p-values only for the primary outcome and report 95 percent confidence intervals without p-values for all secondary outcomes. The confidence intervals have not been adjusted for multiple comparisons and inferences drawn from them may not be reproducible.

Results

Study population

The study population averaged 1.8 hospital admissions in the six months prior to the index admission, (Table 1) compared to less than 0.1 in the general adult Camden population (see supplemental appendix). The study population was 50 percent male; 40 percent were under 55, and 30 percent were over 65; 55 percent were non-Hispanic African American, 30 percent were Hispanic, and 15 percent were non-Hispanic white. Our pre-specified covariates were balanced between treatment and control (Table S2).

Tables S1 and S2 show that three-quarters of the study population were unmarried, one-half had less than a high school degree, and three-fifths reported needing help with mobility. Almost the entire population (95 percent) was not employed, and 40 percent were diagnosed with substance abuse during the index admission. 48 percent had Medicare as their primary payer, and 45 percent had Medicaid as their primary payer.

Program implementation

Table 2 presents measures of program implementation. Ninety-five percent of the treatment group had at least three encounters with program staff after enrollment; on average, a patient received 7.6 home visits, 8.8 phone calls from staff, and was accompanied on 2.5 physician visits. Ninety percent worked with the Coalition for more than 30 days; median program duration was 92 days. The Coalition set ambitious timing goals³⁶: a home visit from program staff within 5 days of arriving home, and a provider visit within 7 days of arriving home; 60 percent met the first goal, 36 percent met the second, and 28 percent met both. Three quarters received both a home visit within 14 days and a provider visit within 60 days.

Receipt of government benefits during the six months post discharge was the one metric of program implementation we observed for both treatment and control groups. Rates of participation in both Temporary Assistance for Needy Families and General Assistance were low and did not significantly change with the intervention; the adjusted difference in

Supplemental Nutrition Assistance Program participation associated with the intervention was 4.6 percentage points (95% CI = 0.5 to 8.6).

Results from randomized evaluation

Table 3 shows results from the randomized evaluation. The 180-day readmission rate was 61.7 percent in the control and 62.3 percent in the treatment group. The intervention had no significant impact on this primary outcome: the adjusted difference in the probability of readmission was 0.82 percentage points higher in the treatment group relative to the control group (95% CI = -5.97 to 7.61; p-value = 0.81). This finding is robust to using multiple imputation to account for missing data (adjusted difference, 0.64 percentage points; 95% CI = -6.12 to 7.40, see Table S6). The intervention also had no impact on any of the secondary outcomes or within any of the pre-specified subgroups (Table 3).

Results for the primary outcome were not sensitive to alternative specifications or measurement over alternative horizons. There were no significant effects of the intervention when the hazard rate of readmission (with either a Cox proportional model or competing risks model accounting for mortality), 180-day mortality, or post-hoc subgroups were analyzed; results differed slightly by hospital of index admission, but the estimates were quite imprecise (Tables S6, S8; Figure S5).

Pre-post analysis

Figure 2 shows average number of admissions per quarter. In both treatment and control groups, admissions rose sharply in the six months prior to the intervention and fell rapidly afterwards. Pre-post analysis within the treatment group is very sensitive to the definition of the pre-period. There was a 38 percentage point *decline* in the probability of a hospital admission in the six months post intervention compared to the six months prior to the intervention, but a 29 percentage point *increase* in the probability of a hospital admission in the six months post intervention compared to the period twelve to eighteen months prior to the intervention (Table S5).

Discussion

In this randomized evaluation of 800 study participants, the Camden Core Model had no significant effect on patients' 180-day readmission rate. The 95 percent confidence intervals rule out a decline in readmission rates of more than 6 percentage points, compared to a control mean of 62 percent; this rules out the 15–45 percent reductions in readmissions in the Medicare population found in randomized evaluations of other care transition programs.^{24–29} The Camden Model targets a different population: younger, with more diverse medical needs, greater social complexity, and much higher healthcare utilization; prior hospital use is nearly twice that in most previous successful care transition programs.

Our results suggest challenges for super-utilizer programs aimed at medically and socially complex populations. This is consistent with randomized evaluations of care management programs of chronically ill, non-Medicare populations that have not found impacts on hospital admissions,^{37,38} although programs like these—which do not focus on the post-discharge transition—have also shown mixed results in a Medicare population.³⁹ It is

possible that care management approaches, designed to connect patients to existing resources, are insufficient for these complex cases. The Coalition's model has continually adapted, and both they and others are exploring models involving more complete redesigns of care provision.^{6,40,41}

Engagement with the program was high (95 percent had at least three encounters) and patients received an intensive intervention (averaging 7.6 home visits), but two program goals on the timing of services—home visit within 5 days and a provider's office visit within 7 days—were achieved less than 30 percent of the time. Challenges in reaching these goals included patients without stable housing or phones, behavioral health complexities, and providers with few available appointments. The difficulties that this pioneering, data-driven organization had in achieving rapid assistance for patients may portend difficulties for achieving it at scale.

Our findings may also reflect fundamental challenges with the strategy of targeting today's super-utilizers: many current high-cost patients will not be high-cost in the future—and this becomes even more pronounced as one goes higher in the cost distribution.^{22,42,43} Moreover, for those with persistent high costs, very little spending may be on potentially preventable hospitalizations.^{42–44}

Such regression to the mean also underscores the importance of rigorous evaluation through randomized trials, as observational evaluations of super-utilizer programs will be prone to find spurious impacts.^{18,22,23} This danger was illustrated in our setting by the similar reduction in readmissions in both the treatment and control groups.

This study has several limitations. It was powered to detect whether this care transition program could achieve reductions in re-admissions comparable to similar programs focused on less-complex patients. But the study was not powered to detect smaller reductions that could be clinically meaningful, nor was it powered to analyze effects within specific subgroups, where there could be differential impacts. The data did not permit evaluation of potential non-tangible benefits, such as improved relationships with providers.⁴⁵ Nor did they allow comparison of outpatient care for treatment and control groups; usual care in Camden was evolving over the study period with multiple other care management programs starting^{46–49} and the Coalition leading a citywide campaign to connect patients to primary care within 7 days of discharge.⁵⁰

Despite these limitations, this study provides rigorous evidence of the impact of a nationally recognized program aimed at super-utilizers of the healthcare system that has been expanded to other cities. The results suggest challenges in reducing readmissions in a medically and socially complex super-utilizer population, as well as the importance of randomized evaluation of interventions that, because they target high-cost patients, likely exhibit substantial regression to the mean in observational studies.

Disclosure forms provided by the authors are available with the full text of this article at [NEJM.org](https://www.nejm.org).

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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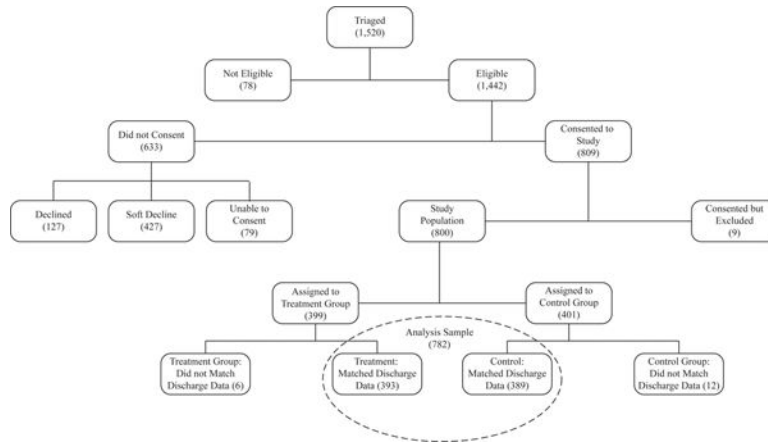


Figure 1: Consort Diagram

Notes: Data are from the Health Information Exchange. “Declined” indicates the patient explicitly said no to the offer of randomization. “Soft Decline” indicates that the patient did not give consent when approached, but did not decline and could be reapproached during future hospitalizations if otherwise eligible. “Unable to Consent” indicates that the patient was discharged (or died) before they were reached or that they were unable to consent for reasons such as being asleep. “Consented but Excluded” includes five patients who consented and later asked to be removed and the last 4 patients enrolled in the study who were excluded to keep the study population at the 800 person target.

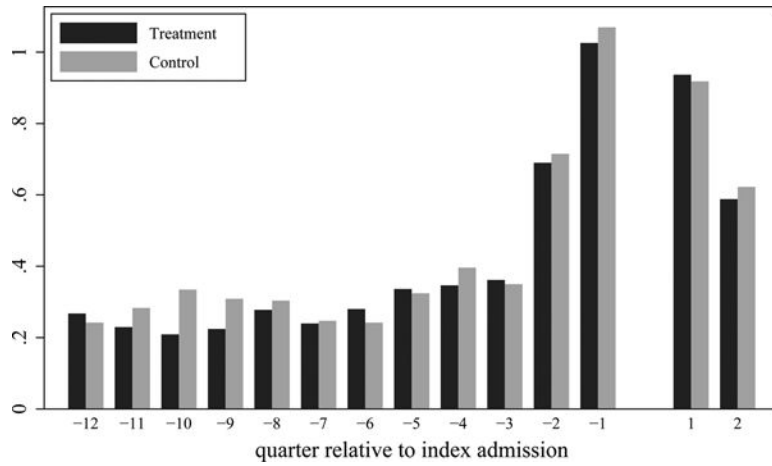


Figure 2: Average Number of Inpatient Admissions per Quarter

Notes: All data are from the hospital discharge data and cover the analysis sample (N=782). Treatment data (N=393) and Control data (N=389) are shown separately. The horizontal axis shows quarters relative to the index admission; quarter 1 is thus the quarter that begins with the discharge date from the index admission, while quarter -1 is the quarter ending the day before the index admission. The index admission is excluded from the figure.

Table 1:

Summary Statistics

	Overall	Treatment	Control
Study Population (N=800):			
Index Admission Match Rate (%)	97.8	98.5	97.0
Analysis Sample (N=782):			
Observations	782	393	389
Male (%)	50.0	52.4	47.6
Age at index admission (%)			
<44	17.1	16.0	18.3
45–64	55.4	55.0	55.8
>65	27.5	29.0	26.0
Race (%)			
African American Non-Hispanic	54.9	57.8	51.9
Hispanic	29.5	26.7	32.4
White Non-Hispanic	15.1	14.8	15.4
Asian / Multiracial / Other	0.5	0.8	0.3
Number of inpatient admissions prior to the index admission			
0–6 months prior	1.75	1.72	1.78
7–12 months prior	0.74	0.74	0.75
Primary payer (%)			
Medicaid	44.6	43.0	46.3
Medicare	48.2	47.6	48.8
Other	7.0	9.2	4.9
Employment status (%)			
Currently Employed	5.5	4.8	6.2
No response	0.5	0.3	0.8
Mental health diagnoses at index admission (%)			
Depression	30.2	32.3	28.0
Substance Abuse	44.0	41.2	46.8

Notes: “Study Population” and “Analysis Sample” are defined in Figure 1. Within “Analysis Sample,” data on gender, age, number of admissions prior to index, and mental health diagnoses are from the hospital discharge data. Race, primary payer, and employment status come from the baseline survey. The analysis sample excluded 18 participants with missing outcome data because they could not be matched to the discharge record for their index admission.

Table 2:

Program Metrics and Benefit Participation

Program Metrics for Treatment Group		Mean	Median	At least 1 encounter (%)	At least 3 encounters (%)
<i>Number of Encounters</i>					
Home visits		7.6	5	88.8	70.7
Phone calls		8.8	5	88.0	65.4
Primary care physician (PCP) and specialist visits		2.5	2	84.7	29.5
Other types of visits		5.7	1	65.1	36.1
Total encounters		28.1	17	98.7	95.2
<i>Length of Intervention (measured from discharge home)</i>		Greater than 30 days (%)	Greater than 90 days (%)	Greater than 180 days (%)	Median number of days
		89.8	50.5	17.0	91.5
<i>Timing of Services (measured from discharge home) (%)</i>					
Camden Coalition home visit within 5 days		58.6			
Camden Coalition home visit within 14 days		83.0			
Visited PCP/Specialist within 7 days		36.0			
Visited PCP/Specialist within 14 days		60.2			
Visited PCP/Specialist within 60 days		83.3			
Both home visit within 5 days and PCP/Specialist visit within 7 days		28.0			
Both home visit within 14 days and PCP/Specialist visit within 60 days		76.1			
Benefit Participation during 6 months post enrollment					
	Control (1)	Treatment (2)	Unadjusted Difference (95% CI) (3)	Adjusted Difference (95% CI) (4)	
Supplemental Nutrition Assistance Program (%)	50.13	58.52	8.4 (1.43 to 15.36)	4.59 (0.52 to 8.65)	
Temporary Assistance for Needy Families (%)	1.03	1.78	0.75 (-0.9 to 2.4)	0.69 (-0.34 to 1.71)	
General Assistance (%)	6.94	6.87	-0.07 (-3.63 to 3.49)	0.68 (-1.82 to 3.18)	

Notes: Data on program metrics are from the Camden Coalition records and consist of the treatment group in the analysis sample (N=393). Data are missing for 4 participants for Timing of Services and 11 participants in Length of Intervention. Data about benefit participation are from the New Jersey Department of Human Services and consist of the analysis sample (N=782). Column (1) shows the mean value for each outcome in the control group, and column (2) shows the mean value for each outcome in the treatment group. Column (3) shows the coefficient (and 95% confidence interval in parentheses) on an indicator for treatment group from an ordinary least squares regression of the outcome with no other covariates. Column (4) shows the coefficient (and 95% confidence interval in parentheses) on an indicator for treatment group from an ordinary least squares regression of the outcome with pre-specified covariates. All confidence intervals are calculated using heteroskedasticity-robust standard errors.

Pre-specified covariates include the dependent variable 0–6 months before the index admission, the dependent variable 7–12 months before the index admission, and indicators for age (in five-year bins), male, African-American non-Hispanic, Hispanic. The covariates are measured in the hospital discharge data except for race which is from the baseline survey.

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Table 3:

Impact of Intervention, 180 days post discharge

	Control Mean (1)	Treatment Mean (2)	Unadjusted Difference (95% CI) (3)	Adjusted Difference (95% CI) (4)
Analysis sample (N=782)				
Any readmission (%)	61.70	62.34	0.64 (-6.17 to 7.46)	0.82 (-5.97 to 7.61)
Number of readmissions	1.54	1.52	-0.02 (-0.29 to 0.26)	0.01 (-0.25 to 0.27)
2 or more readmissions (%)	36.25	36.39	0.14 (-6.61 to 6.89)	0.27 (-6.22 to 6.77)
Days in hospital	9.95	9.36	-0.59 (-2.49 to 1.31)	-0.32 (-2.17 to 1.53)
Hospital charges (\$)	114,768	116,422	1,654 (-25,523 to 28,831)	3,722 (-23,438 to 30,882)
Hospital payments received (\$)	17,650	18,130	480 (-3,613 to 4,573)	680 (-3,415 to 4,775)
Any readmission by subgroup (%)				
2 admissions in prior year (N=336)	52.12	52.63	0.51 (-10.2 to 11.22)	0.78 (-10.35 to 11.91)
3 + admissions in prior year (N=446)	68.75	69.82	1.07 (-7.51 to 9.65)	1.27 (-7.38 to 9.92)
English preferred language (N=638)	63.11	62.61	-0.49 (-8.01 to 7.02)	0.1 (-7.42 to 7.61)
Other preferred language (N=144)	56.25	60.94	4.69 (-11.58 to 20.96)	8.49 (-9.08 to 26.06)

Notes: Data consist of the analysis sample (N=782) and outcomes are measured in the hospital discharge data. Column (1) shows the mean value for each outcome in the control group. Column (2) displays the mean value for each outcome in the treatment group. Column (3) shows the coefficient (and 95% confidence interval in parentheses) on an indicator for treatment group from an ordinary least squares regression of the outcome with no other covariates. Column (4) shows the coefficient (and 95% confidence interval in parentheses) on an indicator for treatment group from an ordinary least squares regression of the outcome with pre-specified covariates. All confidence intervals are calculated using heteroskedasticity-robust standard errors. Pre-specified covariates include the number of admissions 0–6 months before the index admission, the number of admissions 7–12 months before the index admission, and indicators for age (in five-year bins), male, African-American non-Hispanic, Hispanic. The covariates are measured in the hospital discharge data except for race which is from the baseline survey. For three of the outcomes (days in hospital, hospital charges, and hospital payments received), we replace the number of admissions 0–6 months before index admission and 7–12 months before index admission with the values of the dependent variable over those two time periods. The p-value of the primary outcome (any readmission) for the adjusted difference is 0.81.