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Variation in 30-Day Readmission Rates from Inpatient **Rehabilitation Facilities to Acute Care Hospitals**

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Abstract

Objectives: To quantify the rate of readmission from inpatient rehabilitation facilities (IRFs) to acute care hospitals (ACHs) during the first 30 days of rehabilitation stay. To measure variation in 30-day readmission rate across IRFs, and the extent that patient and facility characteristics contribute to this variation.

Design: Retrospective analysis of an administrative database.

Setting and Participants: Adult IRF discharges from 944 US IRFs captured in the Uniform Data System for Medical Rehabilitation database between October 1, 2015 and December 31, 2017.

Methods: Multilevel logistic regression was used to calculate adjusted rates of readmission within 30 days of IRF admission and examine variation in IRF readmission rates, using patient and facility-level variables as predictors.

Results: There were a total of 104,303 ACH readmissions out of a total of 1,102,785 IRFs discharges. The range of 30-day readmission rates to ACHs was 0.0% - 28.9% (mean = 8.7%, standard deviation = 4.4%). The adjusted readmission rate variation narrowed to 2.8%-17.5% (mean = 8.7%, standard deviation = 1.8%). Twelve patient-level and 3 facility-level factors were significantly associated with 30-day readmission from IRF to ACH. A total of 82.4% of the variance in 30-day readmission rate was attributable to the model predictors.

Conclusions and Implications: Fifteen patient and facility factors were significantly associated with 30-day readmission from IRF to ACH and explained the majority of readmission variance. Most of these factors are nonmodifiable from the IRF perspective. These findings highlight that adjusting for these factors is important when comparing readmission rates between IRFs.

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Keywords

Post-acute care; patient readmission; hospital; rehabilitation

Inpatient rehabilitation facilities (IRFs) account for over 500,000 acute care hospital (ACH) discharge dispositions on an annual basis.¹ Patients are required to meet a level of functional impairment and medical stability before they are approved for admission to an IRF.^{2,3} Despite this, medical complications commonly occur after discharge to IRF and approximately 10% of patients will require a readmission from the IRF to an ACH for further medical care within 30 days.⁴ These transfers of care can place patients at risk for service duplication, medical errors and adverse events.^{5,6} From the IRF perspective, readmissions to an ACH are an important quality metric because they disrupt a patient's rehabilitation course and place the patient at risk of losing functional gains achieved during their IRF stay.

Apart from impacting the patient directly, readmissions also result in substantial financial cost to the health care system. According to the Nationwide Readmissions Database in 2016 each hospital readmission costs an estimated \$14,400.⁷ The total financial impact of 30-day hospital readmissions in the United States in 2011 was \$41.3 billion.⁸ Numerous policies aimed at curbing rising health care costs have targeted hospital readmissions as a key area for improving clinical care coordination and achieving potential savings. Beginning in 2013, Medicare began to impose penalties on hospitals for high rates of readmissions for certain highly prevalent diagnoses as part of the Hospital Readmissions Reduction Program. In 2017, these penalties imposed by Medicare cost hospitals more than \$500 million.⁹ Motivated by the risk of payment reduction and the negative impact on reputation, hospitals have a strong incentive to reduce readmissions.

Among the strategies implemented toward the goal of readmission reduction, hospitals have attempted to increase patient education, multidisciplinary team care, and coordination of post-acute services.¹⁰ However, despite modest progress, readmission rates remain high. According to the 2019 US Agency for Healthcare Research and Quality statistical brief, all-cause 30-day readmissions only decreased from 14.2% in 2010 to 13.9% in 2016.⁷ Several studies have analyzed causes for these readmissions, and have concluded that a large percentage of readmissions are either unavoidable or caused by nonmodifiable patient and facility characteristics.^{11,12} However, no studies have provided a national assessment of readmission variability from IRFs across the wide spectrum of common rehabilitation impairment diagnoses.

Therefore, the purpose of this study was (1) to quantify readmissions from US IRFs to ACHs during the first 30 days of the rehabilitation admission across a wide array of common rehabilitation impairment diagnoses; (2) to measure the variation in readmission rates between IRFs; and (3) to quantify the extent that patient and facility characteristics account for this variation.

Methods

Study Design

This study was a retrospective review of administrative data.

Data Source

Data was obtained from the Uniform Data System for Medical Rehabilitation (UDSMR). The UDSMR database includes demographic and medical data, as well as information regarding facility characteristics, collected from the IRF-Patient Assessment Instrument (IRF-PAI) from IRFs in the United States.¹³ The UDSMR includes approximately 70% of US IRFs. The IRF-PAI is an assessment instrument required for Medicare payments that is widely used by IRF providers for quality measures.¹⁴

Facilities

All US IRFs within the UDSMR database with at least 60 total patient discharges between the period of October 1, 2015 and December 31, 2017 were included in this study. The minimum criteria of 60 discharges during the selected time period was chosen based on similar criteria in prior work¹⁵ and to exclude very low volume facilities.

Study Population

IRF discharges of adult patients age 18 years or older were included in this study. Sixteen diagnosis impairment groups (Table 1) were included in this study, as defined by the IRF-PAI manual.¹⁶ The only IRF-PAI impairment group not included in this study was the developmental disability impairment group because of small size (~0.001% of the cohort).

Study Variables

Based on similar prior work, patient and facility characteristics available in the UDSMR database that were thought to have a potential effect on risk of readmission from IRF to ACH within the first 30 days of rehabilitation stay were included as study variables.^{15,17–19}

Patient Characteristics

Patient characteristics included sex, age, duration of impairment (days between impairment onset and IRF admission), race/ethnicity (Caucasian, African American, Latino/Hispanic, Asian, or other, which included no race, multiracial, and other race), marital status (married, not married), living status (living alone, living with others), primary payer source (Medicare, Medicaid, commercial insurance, unreimbursed, worker's compensation, or other), presence of dysphagia or pneumonia on admission, presence of other comorbidities (as measured by the Elixhauser Comorbidity Index²⁰), weekend admission status (defined as admission to the IRF on a Friday, Saturday, or Sunday²¹), and admission cognitive and motor functional independence measure (FIM) scores.¹⁶ The FIM is a validated instrument that measures function using 18 items subcategorized into motor (13 items) and cognitive (5 items) domains.²² It was developed for tracking rehabilitation outcomes and was subsequently incorporated into the IRF-PAI for use in Medicare's payment system for IRFs. Age, duration

of impairment, Elixhauser Comorbidity Index, and motor and cognitive FIM were treated as quantitative variables. All other variables were treated as categorical variables.

Facility Characteristics

Facility type was defined as either freestanding or within an ACH. Similar to prior studies, geographic location was divided by Center for Medicare and Medicaid Services regions designation as follows: Eastern (regions I-IV), Central (regions V-VIII), and Western (regions IX and X).¹⁵ A facility was consider accredited by the Commission on Accreditation of Rehabilitation Facilities (CARF) if it received accreditation at any time over the study period. Facility size was defined by the number of operating beds. Other examined facility characteristics included mean admission cognitive and motor FIM (defined as mean admission score of all patients at a given IRF) and mean duration of impairment (defined as mean number of days from impairment onset to IRF admission at a given IRF).

Outcome Measures

The primary outcome was readmission from IRF to an ACH within 30 days of IRF admission. This outcome was compared with all other discharges from the IRF during the study period, including discharges to ACHs outside of the initial 30-day IRF stay and discharges to home or to other post-acute care facilities (such as skilled nursing facilities) at any time during the study period.

Statistical Analysis

Stata v 16.1 (Stata Statistical Software: Release 16; StataCorp LLC, College Station, TX, USA) was used to complete the statistical analyses. Patient and facility characteristics were compared between 30-day ACH readmissions and all other discharges using 2-sample t-tests for quantitative variables and χ^2 tests for categorical variables. An adjusted multilevel logistic regression model was built to model rate of 30-day readmission from IRF to ACH. All patient and facility characteristics described above that were statistically significant model predictors of 30-day readmission to ACH were included in the final model. The relationships between patient and facility variables and the outcome of 30-day readmission were assessed graphically for potential nonlinear relationships, and quadratic and piece-wise treatment of variables was used when appropriate. The final model was then used to calculate adjusted rates of 30-day readmission to ACH for the study IRFs. IRF rank (by readmission rate) vs 30-day readmission rate was plotted for the raw (unadjusted) data and the modeled (adjusted) data. The variance of readmission rates was calculated for the unadjusted (v_1) and adjusted (v_2) distributions then the percent of variance attributed to the model variables was calculated as $[1-(v_2/v_1)] \times 100$. For all statistical tests, a P value of less than .05 was considered statistically significant.

Results

A total of 1,102,785 discharges from 944 IRFs met inclusion criteria for this study. Data from 12 IRFs were not included due to the facilities having fewer than 60 discharges during the study period. Of the 16 impairment groups, stroke was most common (23.5%), followed by orthopedic conditions (22.6%), and neurologic conditions (13.2%, Table 1). A total of

104,303 (9.5%) of all discharges from the IRFs were readmissions to ACHs within 30 days. The highest 30-day readmission rates to ACHs occurred among the medically complex group (13.6%), whereas the lowest rates occurred among orthopedic conditions (5.7%).

Patients readmitted within 30 days were older, more likely to be male, have lower admission FIM cognitive and motor scores, have more comorbid medical conditions, and be admitted over the weekend compared with all other IRF discharges. Thirty-day readmissions had the following facility factors compared with all other discharges: larger facility, freestanding status, and less CARF accreditation (Table 2).

In the final multilevel logistic regression model, 12 patient-level and 3 facility-level factors were significantly associated with 30-day readmission from IRF to ACH. Odds of 30-day readmission rate to ACH was greater for patients who were male, married, had more comorbidities, and who were admitted on the weekend. Odds of 30-day readmission to ACH was greater for freestanding hospitals and lower facility mean admission motor FIM score (Table 3). The following were not included in the final logistic regression model, as they were not found to be statistically significant predictors: living status, primary payer, geographic region, CARF accreditation, and facility size.

Across the studied 944 IRF facilities, the range of 30-day readmission rates to ACHs was 0.0%-28.9% (mean = 8.7%, standard deviation = 4.4%). The adjusted readmission rate variation narrowed to 2.8%-17.5% (mean = 8.7%, standard deviation = 1.8%). This is depicted by the flattening of the curve in Figure 1. The model variables accounted for 82.4% of the variance in 30-day readmission rate of the unadjusted sample. Of the study IRFs, 55.0% (n = 519) had adjusted 30-day readmission rates with 95% confidence intervals crossing the mean readmission rate across facilities. A total of 22.0% (n = 208) fell below the mean and 23% (n = 217) were above the mean.

Discussion

This retrospective analysis is one of the largest to date examining readmission variation from IRFs to ACHs. It includes patients from a diverse group of rehabilitation impairment diagnoses across all regions of the United States and is not limited to a particular payer population. Thirty-day readmission rates to ACHs were found to vary greatly across the 944 IRFs (0.0%–28.9%). The adjusted readmission rate variation narrowed to 2.8%–17.5%. Ramey et al¹⁵ also observed a substantial decrease in readmission rate variation following adjustment for patient and facility characteristics in a smaller cohort of rehabilitation patients with the medically complex impairment diagnosis (from 0%–44.4% to 6.9%–21.9% following adjustment for 9 patient and facility factors). In our larger cohort of multiple IRF impairment diagnoses, significant patient and facility factors accounted for 82.4% of the observed variance in 30-day readmission rates. This is similar in magnitude to prior analyses of general readmissions to ACHs among Medicare patients, in which more than one-half of readmission variability has been attributed to patient and facility characteristics.^{18,19} It is important to understand the factors that contribute to need for ACH readmission because there are known risks of transfers-of-care (provider discontinuity, medication errors, loss of

functional gains, etc).⁶ By understanding which factors are potentially modifiable, quality efforts can be formulated to reduce avoidable transfers and readmissions from IRFs.

In this study, a number of patient characteristics were found to impact risk of 30-day readmission from IRFs to ACHs. Most of these, such as impairment diagnosis, functional status on admission, duration of impairment, medical comorbidities, and demographic factors, are nonmodifiable factors from the standpoint of the IRF. This study agrees with multiple prior works that have shown association of worse mobility at IRF admission with increased risk of 30-day acute hospital readmission.^{23–25} Kumar et al found that in the Medicare stroke population greater duration of physical therapy received at the ACH prior to discharge correlated with lower rate of 30-day hospital readmission.²⁶ These findings may reflect more than just correlation, as there is some data to suggest that early hospital-based mobility interventions in the ACH setting can reduce readmissions.^{27,28} Early hospital-based therapies may have a protective role against hospital readmission by optimizing predischarge mobility and functional status. Other significant modifiable factors in this study, such as day of admission, could also be a target for hospital quality improvement efforts. Echoing the findings of Shih et al,²¹ in this cohort weekend IRF admissions were at higher risk for 30-day readmission to ACH. A reasonable risk-reduction strategy could be to avoid weekend admissions for patient with other high-risk factors.

Among the examined facility factors, CARF accreditation, arguably the only modifiable facility factor examined in this study, was not found to be a statistically significant predictor of risk of 30-day readmission to ACH in the final model. Although CARF accreditation is not required, CARF is the industry gold standard and gaining certification requires display of strict quality standards and performance improvement processes by the institution.²⁹ Although one might hypothesize that accreditation might reduce risk of 30-day readmission to ACH, this was not observed in this study cohort. The facility characteristics that did significantly influence risk for 30-day readmission (freestanding hospital status, mean facility admission motor and cognitive FIM, and mean duration of impairment on admission) are nonmodifiable from the standpoint of the IRF. Higher readmission rates observed for freestanding IRFs vs in-hospital IRFs may reflect the decreased access to resources such as diagnostic tests and specialized physicians at freestanding IRFs. The fact that these factors are largely outside the control of the IRF highlights the importance of correcting for these factors if 30-day readmission to ACH is used as a quality metric to compare IRF facilities.

Although this study specifically examined IRFs, it is important to note that in the United States skilled nursing facilities (SNFs) are an even more common discharge destination from ACHs.³⁰ The 2 have a number of differences, which may influence readmission risk. While patients discharged to IRFs must be able to tolerate and benefit from 3 hours of rehabilitative therapy 5 days a week, this is not required for discharge to SNFs. From a staffing standpoint, IRFs are required to have 24-hour nursing availability and rehabilitation physician-led interdisciplinary treatment, with least \times 3 weekly in-person physician visits. In contrast, SNFs are not necessarily required to have on-site nursing around the clock, and services are not necessarily supervised by a rehabilitation physician. Because of possible need for greater staffing and resources, patients who require more complex nursing care (such as patients with ulcers, dysphagia, or incontinence) or more frequent laboratory

monitoring may be more likely to go from ACH to IRFs than SNFs.³¹ There is also some difference in patient age between IRFs and SNFs, reflecting the difference in common diagnoses; generally, more traumatic diagnoses are treated at IRF (such as spinal cord injury, traumatic brain injury, orthopedic trauma), which translates to a younger population than that seen at SNFs. An additional difference is that length of stay is generally shorter in IRFs than SNFs; per a Medicare Payment Advisory Commission (MedPAC) analysis, average length of stay for stroke was 15 vs 25 days at IRFs vs SNFs, and for hip/femur orthopedic procedures was 14 vs 32 days.³¹ This may be partly influenced by different reimbursement systems. While SNFs are paid by Medicare on a per day basis, IRFs receive a bundled "per discharge" payment based on the patient's rehabilitation diagnosis and comorbidities.³⁰ Furthermore, a patient copay begins at day 21 of SNF stay for Medicare patients. In terms of ACH readmission rates, unadjusted 30-day ACH readmission rates tend to be higher for SNFs than for IRFs (15.3% vs 11.1% for stroke, 11.3% vs 8.4% for hip and femur orthopedic procedures).³¹ Possible drivers of these higher readmission rates may be greater comorbidities in the SNF population,³⁰ or decreased access to physicians and diagnostic testing (laboratories, imaging, etc) at SNFs compared with IRFs.

Understanding the risk factors for increased readmission rates from the IRF setting has critically important implications for performance measurement and quality benchmarking for IRF providers and health systems more broadly. Starting in 2016, 30-day hospital readmissions for the Medicare population began to be reported publicly on the IRF Compare website for all federally licensed IRFs.³² More recently, US News and World Report and Newsweek have proposed incorporating 30-day hospital readmissions rates from the IRF Compare site into its rankings methodologies to guide consumer decision-making.^{33,34} Furthermore, the MedPAC has highlighted 30-day readmissions for inclusion in a proposed value-based incentive program for providers across the post-acute care spectrum, including IRFs.³⁵

Despite expected trends toward public reporting of performance metrics,³⁶ such reporting efforts have proved controversial and underwhelming in the past.^{37–39} Standardized reporting methods and risk-adjustment of outcomes are consistently recommended as measures that can help increase provider acceptance and broad credibility to both consumers and providers in public reporting frameworks.⁴⁰ Given the heterogeneity of IRF populations nationally, it will be crucial for IRF providers and consumers to understand the facility-based and patient-based risk factors that may contribute to readmissions as an outcome.

This study has a several limitations. One limitation is that this study only captures patients directly discharged from IRFs to ACHs within 30 days. It does not capture those who are discharged to the community, then readmitted to an ACH within 30 days of initial acute care discharge. This is another important population from the IRF standpoint that has been examined by previous investigators.^{17,41} In addition, this dataset does not distinguish between planned and unplanned ACH readmissions. Nor does it include detailed characteristics about the medical or rehabilitation treatments received by patients or examine readmission risk of individual medical comorbidities. On the other hand, medical comorbidities may not always correlate with functional status, which may be a stronger predictor of readmission risk. It also lacks some detailed information about the IRFs and

ACHs (such as overall ACH size). Furthermore, while the UDSMR database captures the majority of US IRFs, it is not comprehensive, as participation in the UDSMR database is voluntary. Therefore, we are unable to know whether results from this subset of IRFs are fully generalizable to all US IRFs. Finally, because the UDSMR is a deidentified dataset, the analysis is unable to identify or correct for multiple readmissions of the same patient (an individual may be represented more than once in the dataset for separate readmission events). For this reason, descriptive statistics describe readmissions rather than patients in this analysis. Despite these limitations, this is one of the largest studies to-date examining variations in 30-day readmission to ACH from IRF among a diverse population from a geographic, medical, and payer standpoint. Therefore, it contributes to understanding the magnitude that patient and facility factors explain the observed variation across the US in readmissions from IRFs.

Conclusions and Implications

Large variation in rates of 30-day readmission to ACHs was observed in this nationwide sample of IRFs. Fifteen patient and facility factors were significantly associated with 30-day readmission from IRF to ACH and explained the majority of readmission variance. Most of these factors are nonmodifiable from the IRF perspective. This highlights that reporting of unadjusted readmission rates may be misleading, and that adjusting for these patient and facility factors is important when comparing readmission rates between facilities. Future research should continue to examine factors that contribute to variability across IRF facilities to guide quality improvement efforts and decrease potentially avoidable transfers-of-care.

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Table 1

Study Population by Impairment Diagnosis and Discharge Status

Impairment Diagnosis	Readmitted to ACHs from IRF within 30 d, n (% of Impairment Group)	All Other Discharges from IRF, n (% of Impairment Group)	Total (% of Total IRF Discharges)
Medically complex conditions	1066 (13.6)	6759 (86.4)	7825 (0.7)
Pulmonary disorders	2114 (12.9)	14,250 (87.1)	16,364 (1.5)
Cardiac disorders	6412 (12.5)	44,711 (87.5)	51,123 (4.6)
Debility	11,468 (11.9)	84,950 (88.1)	96,418 (8.7)
Brain dysfunction	14,077 (11.4)	109,828 (88.6)	123,905 (11.2)
Amputation	3917 (11.5)	30,040 (88.5)	33,957 (3.1)
Neurologic conditions	16,494~(11.3)	128,891 (88.7)	145,385 (13.2)
Other disabling impairments	1144 (11.0)	9249 (89.0)	10,393 (0.9)
Spinal cord dysfunction	6692 (10.3)	58,507 (89.7)	65,199 (5.9)
Burns	149 (9.7)	1392 (90.3)	1541 (0.1)
Stroke	23,275 (9.0)	236,207 (91.0)	259,482 (23.5)
Congenital deformities	33 (8.8)	342 (91.2)	375 (0.0003)
Major multiple trauma	2612 (8.0)	30,190 (92.0)	32,802 (3.0)
Arthritis	347 (7.4)	4332 (92.6)	4679 (0.4)
Pain syndromes	297 (7.4)	3712 (92.6)	4009 (0.4)
Orthopedic conditions	14,206 (5.7)	235,122 (94.3)	249,328 (22.6)
Total	104,303 (9.5)	998,482 (90.5)	$1,102,785\ (100.0)$

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Table 2

Patient and Facility Characteristics of Discharge Groups

	Readmitted to ACHs from IRF within 30 d	All Other Discharges from IRF	P Value	All Discharges from IRF	Missing Data, n (%)
Patient characteristics					
Age, mean y (SD)	69.2 (14.5)	68.7 (15.2)	<.001	68.8 (15.1)	0 (0.0)
Male, n (%)	56,608 (54.3)	481,840 (48.3)	<.001	538,448 (48.8)	207 (0.02)
Race/ethnicity, n (%)			<.001		33,854 (3.1)
Caucasian	78,967 (77.9)	761,337 (78.7)		840,304 (78.6)	
African American	13,616 (13.4)	121,341 (12.5)		134,957 (12.6)	
Latino/Hispanic	5874 (5.8)	56,120 (5.8)		61,994 (5.8)	
Asian	1725 (1.7)	18,175 (1.9)		19,900 (1.9)	
Other	1138 (1.1)	10,638 (1.1)		11,776 (1.1)	
Married, n (%)	51,267 (50.6)	459,576 (47.4)	<.001	510,843 (47.7)	32,107 (2.9)
Living alone, n (%)	24,958 (24.3)	269,440 (27.3)	<.001	294,398 (27.0)	14,189 (1.3)
Primary payer source, n (%)			<.001		24,041 (2.2)
Medicare	76,755 (75.2)	720,889 (73.8)		797,644 (73.9)	
Medicaid	6271 (6.1)	62,419 (6.4)		68,690 (6.4)	
Commercial	15,379 (15.1)	148,336 (15.2)		163,715 (15.2)	
Unreimbursed	1099 (1.1)	13,599 (1.4)		14,698 (1.4)	
Worker's compensation	530 (0.5)	8613 (0.9)		9143 (0.8)	
Other	2019 (2.0)	22,835 (2.3)		24,854 (2.3)	
Duration of impairment, mean d (SD)	14.1 (21.3)	11.0(20.8)	<.001	11.3 (20.9)	22,982 (2.1)
FIM cognitive admission, mean (SD)	20.3 (7.7)	22.5 (7.1)	<.001	22.3 (7.2)	0 (0.0)
FIM motor admission, mean (SD)	30.4 (11.8)	37.1 (12.6)	<.001	36.5 (12.7)	0 (0.0)
Elixhauser comorbidity Index, mean (SD)	7.8 (7.6)	5.7 (7.1)	<.001	5.9 (7.2)	0 (0.0)
Dysphagia, n (%)	22,458 (21.5)	162,160 (16.2)	<.001	184,618 (16.7)	0 (0.0)
Pneumonia, n (%)	12,264 (11.8)	77,888 (7.8)	<.001	90,152 (8.2)	0 (0.0)
Admitted Friday–Sunday, n (%)	33,305 (31.9)	314,182 (31.5)	.002	347,487 (31.5)	0 (0.0)
IRF characteristics					
Operating beds, mean (SD)	53 (37)	50 (35)	<.001	51 (35)	255 (0.02)
Facility type, n (%)			<.001		0(0.0)

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	Readmitted to ACHs from IRF within 30 d	All Other Discharges from IRF	P Value	All Discharges from IRF	Missing Data, n (%)
Freestanding	55,857 (53.6)	490,208~(49.1)		546,065 (49.5)	
Within ACH	48,446 (46.4)	508,274 (50.9)		556,720 (50.5)	
Facility region, n (%)			<.001		0 (0.0)
East	51,626 (49.5)	463,780 (46.4)		515,406 (46.7)	
Central	40,985 (39.3)	411,744 (41.2)		452,729 (41.1)	
West	11,692 (11.2)	122,958 (12.3)		134,650 (12.2)	
CARF accredited, n (%)	33,032 (31.7)	331,207 (33.2)	<.001	364,239 (33.0)	0 (0.0)
SD, standard deviation.					

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Table 3

Multilevel Logistic Regression Model Examining Patient and Facility Characteristics Associated with 30-Day Readmission to Acute Care Hospitals

	Odds Kalio	Standard Error	r value	95% Conndence Interval
Patient characteristics				
Age	0.99849	0.00027	<.001	0.99796-0.99902
Female	0.87515	0.00640	<.001	0.86270-0.88778
Race/ethnicity *				
African American	1.02137	0.01143	.059	0.99922 - 1.04402
Latino/Hispanic	0.97522	0.01622	.131	0.94395 - 1.00753
Asian	0.93097	0.02592	.010	0.88153-0.98319
Other	0.98559	0.03369	.671	0.92173-1.05388
Elixhauser comorbidity index	1.02454	0.00051	<.001	1.02354 - 1.02554
Married	1.05348	0.00764	<.001	1.03863 - 1.06856
Dysphagia	0.86300	0.00864	<.001	0.84624 - 0.88009
Pneumonia	1.15869	0.01317	<.001	1.13316-1.18480
Days since impairment onset, 10 d	1 (omitted due	e to collinearity)		
Days since impairment onset, 10–190 d	1.01779	0.00047	<.001	1.01687-1.01872
Days since impairment onset, 10-190 d, squared	0.99987	4.12e-06	<.001	0.99987-0.99988
Days since impairment onset, 190 d	1.00637	0.00093	<.001	1.00455 - 1.00820
Admission motor FIM	0.95560	0.00037	<.001	0.95487-0.95632
Admission motor FIM, squared	0.99986	0.00002	<.001	0.99982-0.99991
Admission cognitive FIM	0.99619	0.00071	<.001	0.99479-0.99759
Admission cognitive FIM, squared	1.00077	0.00006	<.001	1.00065 - 1.00089
Admitted Friday–Sunday	1.03570	0.00776	<.001	1.02060-1.05102
Impairment group *				
Brain dysfunction	1.37499	0.02753	<.001	1.32208-1.43002
Neurologic conditions	1.37909	0.02889	<.001	1.32362-1.43689
Spinal cord dysfunction	1.23158	0.02881	<.001	1.17639–1.28937
Amputation	1.75687	0.04580	<.001	1.66936 - 1.84898
Arthritis	1.12919	0.07256	.059	0.99556-1.28076
Pain syndromes	1.23522	0.08796	.003	1.07430-1.42023

	Odds Ratio	Standard Error	P Value	95% Confidence Interval
Orthopedic conditions	0.87141	0.01776	<.001	0.83730-0.90692
Cardiac disorders	2.04617	0.04859	<.001	1.95312-2.14366
Pulmonary disorders	2.09954	0.06810	<.001	1.97022 - 2.23735
Burns	1.12702	0.11266	.232	0.92649-1.37095
Congenital deformities	0.72475	0.18602	.210	0.43825–1.19857
Other disabling impairments	1.72285	0.07060	<.001	1.58988 - 1.86693
Major multiple trauma	0.95384	0.02907	.121	0.89854-1.01254
Debility	1.70522	0.03621	<.001	1.63571-1.77769
Medically complex conditions	1.99545	0.09067	<.001	1.82542 - 2.18131
Facility characteristics				
Freestanding hospital	1.14593	0.01528	<.001	1.11637-1.17628
Mean admission motor FIM	0.98078	0.00163	<.001	0.97759–0.98399
Mean admission cognitive FIM	1.00619	0.00273	.023	1.00084 - 1.01156
Mean days since impairment	1.00199	0.00125	.111	0.99954-1.00444

All variables were treated as linear predictors except admission motor FIM, admission cognitive FIM, which were treated as quadratic variables, and days since impairment onset, which was treated as a piece-wise variable with linear and quadratic components, based on fit.

Harrell C statistic for final model: 0.71.

* For race, Caucasian was set as the reference group (odds ratio = 1). For impairment group, stroke was set as the reference group (odds ratio = 1).