

ORIGINAL ARTICLE

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Prior hospital admission predicts thirty-day hospital readmission for heart failure patients

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Abstract

Background: Hospital readmission is a significant health burden. More than 20% of heart failure (HF) patients are readmitted within 30 days of discharge leading to billions of dollars in health care expenditures. However, the role of prior hospital admissions to predict 30-day readmission for HF patients is not fully understood.

Methods: We retrospectively analyzed HF hospitalization data for 4 years at a single medical center. Association between prior admission and 30-day readmission after HF hospitalization was assessed using a multivariate logistic regression model.

Results: A total of 1,999 patients with index HF hospitalizations were identified, and 366 of them (18%) were readmitted within 30 days. The rate of readmission was 14%, 20%, and 33% in patients with 0, 1, \geq 2 prior admissions. Patients with one prior admission had a 50% higher risk (confidence interval [CI] 1.10–2.05, p = 0.011) for readmission, while those with \geq 2 prior admissions had a more than 3-fold increase in readmission (CI 2.27–4.09, p < 0.001), after adjustments for relevant clinical covariates. Prior hospital admission provided incremental value in predicting readmissions, shown by the significant improvement in the readmission predictive model (C-statistics increased from 0.57 to 0.63). However, neither the length of stay nor recency of prior admission was a significant factor in predicting readmissions.

Conclusions: Hospital admission prior to an index HF hospitalization is associated with a significantly increased risk for 30-day hospital readmission and could be used to identify patients at high-risk for readmission and potentially target interventions to reduce the risk of readmission for these patients. (Cardiol J 2016; 23, 2: 155–162)

Key words: heart failure, readmissions, prior admissions, predictive model

Introduction

Over 5 million Americans suffer from heart failure (HF), and the medical treatment of HF is associated with significant costs [1, 2]. Importantly, 24% of the patients discharged with a primary diagnosis of HF are readmitted to the hospital

within 30 days, representing a significant public health burden [3–5].

There have been efforts made to predict 30-day hospital readmission by the Centers for Medicare and Medicaid Services [3] and other studies [3, 6–9], some of which include data on prior hospital admissions. However, most of these

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studies were conducted with Medicare patients, and the predictive accuracy of such models may be suboptimal in other cohorts. Furthermore, it is not known if the length of stay or recency of the prior admission plays a role in predicting readmission rates. Thus, studies regarding the utility of prior admissions to predict 30-day hospital readmission in "all-comers", and further exploration of characteristics of prior admission in predicting readmission are warranted.

In the present analysis, we aimed 1) to evaluate whether prior hospital admissions predict 30-day hospital readmission after an index admission for HF in a real-life cohort, 2) to assess whether the length of stay during prior admission or recency of prior admission plays a role in readmission, and 3) to identify whether information on prior hospital admissions improves the prediction model for 30-day hospital readmission.

Methods

Study population

This is a retrospective analysis of hospital admission data in HF patients with an index hospitalization for HF, using data for Strong Memorial Hospital from the Office of Clinical Practice Evaluation at the University of Rochester Medical Center, Rochester, NY, USA.

Patients discharged from the hospital with a primary diagnosis of HF were identified (ICD-9 codes starting with 428., plus 402.01, 402.11, 402.91, 404.01, 404.03, 404.11, 404.13, 404.91, 404.93) between January 1, 2007 and June 30, 2011. Hospitalizations of individuals under the age of 18 were excluded from this analysis. Records on hospital admissions between January 1, 2007 and December 31, 2007 were used to derive the number of hospital admissions before January 1, 2008. Hospitalizations with in-hospital death were excluded from the present analysis. In patients with multiple hospitalizations, one index admission was selected at random for analysis, to avoid selection bias.

Definitions and endpoints

An index admission for each patient was defined as an admission with the primary diagnosis of HF after January 1, 2008. The variable of interest in this study was the frequency of prior admissions, defined as the number of admissions due to any cause in the 365 days prior to their index admission. We considered any admissions vs. no admission, as well as the effect of 0, 1, and 2 or

more admissions, on 30-day hospital readmission rates, as a continuous covariate.

The outcome of our study was 30-day hospital readmission for any cause, defined as a hospitalization for any cause within 30 days after discharge from the index admission for HF.

The study was approved by the University of Rochester Research Subjects Review Board.

Echocardiography data

Results from 2-dimensional transthoracic echocardiography were obtained during the index admission for HF when available. Left ventricular (LV) volumes were measured by Simpson's disk method in the apical 4-chamber view and LV ejection fraction (LVEF) was calculated according to the established American Society of Echocardiography protocols [10].

Statistical analysis

Continuous variables were expressed as mean and standard deviations, categorical variables were summarized as frequencies and percentages. Baseline clinical characteristics were compared between patients with vs. without prior hospital admission in the year before the index admission for HF using the χ^2 -test for dichotomous variables and the Kruskal-Wallis test for continuous variables, as appropriate.

Multivariate logistic regression analyses were performed to evaluate the risk of 30-day hospital readmissions by prior hospital admissions. Candidate variables in the model were pre-specified during prior literature review, and chosen if they showed significant associations (p \leq 0.10) with the exposure or the outcome in univariate analyses. There were 29 variables (15 diagnoses) found to have a significant (p < 0.10) association with 30-day hospital readmission after an index HF admission in the univariate analysis (Supplementary Table 1); these were considered variables for the multivariate logistic regression.

Multivariate logistic regression models were adjusted for relevant clinical covariates selected from the candidate covariates using best subset model regression, forcing age, race, and LVEF in the model. After assessing collinearity among these variables, the model was adjusted for age, race, and LVEF (dichotomized as low LVEF < 40% vs. relatively preserved LVEF $\ge 40\%$). We then included prior hospital admissions, the parameter of interest, as a continuous and dichotomized covariate. Covariate effects are reported as odds ratios (OR) and 95% confidence intervals (CI). We

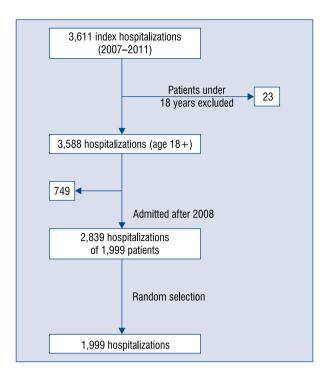


Figure 1. Flowchart of the study population.

reported the overall discriminative value with the receiver-operating characteristic (ROC) curves and results of the C-statistics.

All statistical tests were 2-sided, and a p-value < 0.05 was considered statistically significant. All statistical analyses were performed using SAS version 9.3 (SAS, Cary, NC). The statistical analyses were performed by the first author of the manuscript.

Results

There were a total of 3,611 hospital discharges of patients with the primary diagnosis of HF between January 1, 2007 and June 30, 2011. Twenty-three records were omitted from patients less than 18 years old. In the remaining cohort of 3,588 patients, 749 discharges from the first year were used to calculate prior hospital admission rates for admissions in 2008, as illustrated in Figure 1. Thus, the study population consisted of 1,999 patients with 2,839 HF hospitalizations.

Baseline clinical characteristics

Clinical characteristics of the patients at index HF admission are shown in Table 1. The mean age of the study population was 68 years. A total of 824 (41%) of the patients were women, and 74% were white. Common comorbidities were hypertension (36%), dyslipidemia (42%), ischemic heart disease

(36%), and atrial fibrillation (39%). Patients with prior admission were less often white, more often admitted to the emergency room, and they had more comorbidity, such as atrial fibrillation and other non-cardiovascular diagnoses.

Univariate predictors of 30-day hospital readmission

There were 366 (18%) patients with hospital readmission within 30 days after the index HF admission. Patients with higher risk of mortality and severity of illness scores, a greater number of diagnoses, and longer length of stay during their index admission for HF had a significantly higher frequency of 30-day hospital readmission for any cause. Age, gender, marital status, and discharge disposition did not show differences among patients readmitted to the hospital within 30-days vs. those who were not (Supplementary Table 1).

Prior hospital admission predicts 30-day hospital readmission

The rate of hospital readmission 30 days after an index HF admission among patients with no prior admission was 14%, compared to 26% in those with prior admissions (p < 0.001). Figure 2 illustrates the association between the rate of 30-day hospital readmission and the number of hospital admissions in the year prior to the index HF admission. There were 19.6% of patients readmitted within 30-days among those with 1 prior admission, 32.3% among those with 2 prior admissions, and 34.4% among those with 3 or more prior admission. Because patients with 2 prior admissions and 3 or more prior admissions had similar rates of 30-day readmission, we combined these two groups.

Our results were confirmed in the multivariate logistic regression model, patients with any prior admission had a 31% higher risk (p < 0.001) of 30-day hospital readmission following an index HF admission (Table 2). When we assessed the risk of 30-day hospital readmission by the number of prior admissions, we found that patients with 1 prior admission had a 47% increased odds of readmission within 30-days (p = 0.015); while those with 2 or more prior admissions showed an almost 3-fold increased likelihood of 30-day readmission compared to those with no prior admissions (p < 0.001; Table 3). Compared to other predictors, prior admissions had the highest χ^2 values, thus, associated with the best predictive value in our models.

The ROC curves showed improvement in the C-statistics from 0.57 to 0.62 including prior

Table 1. Clinical characteristics of the study population at the time of index admission, and stratified by prior hospital admissions.

Parameters	All patients	Patients with prior hospital admission (A)	Patients without prior hospital admission (B)	P-value between A and B			
Number of patients	1,999	700	1,299				
Demographics	Demographics						
Age [years]	68.2 ± 15.6	68.2 ± 15.4	68.2 ± 15.8	0.916			
Female gender	824 (41)	307 (44)	517 (40)	0.079			
White race	1,485 (74)	487 (70)	998 (77)	< 0.001			
Married	920 (46)	313 (45)	607 (47)	0.388			
Baseline risk scores							
Admission type (emergency)	1,720 (86)	628 (90)	1,092 (84)	< 0.001			
Risk of mortality* (major or extreme)	950 (48)	344 (49)	606 (47)	0.316			
Severity of illness** (major or extreme)	1,366 (68)	478 (68)	888 (69)	0.896			
Characteristics of index heart failure adm	Characteristics of index heart failure admission						
Number of diagnoses on admission	15.8 ± 6.1	17.2 ± 5.9	15.1 ± 6.1	< 0.001			
Length of stay [days]	7.9 ± 15.2	8.8 ± 18.2	7.5 ± 13.2	0.051			
Disposition to home	1,706 (85)	576 (82)	1,130 (87)	0.005			
Readmission in 30-days	366 (18)	181 (26)	185 (14)	< 0.001			
Co-morbidities at index heart failure adm	ission						
Hypertension	714 (36)	219 (31)	495 (38)	0.002			
Hyperlipidemia	847 (42)	301 (43)	546 (42)	0.676			
Ischemic heart disease	718 (36)	147 (21)	270 (21)	0.910			
Left bundle branch block	122 (6)	28 (4)	94 (7)	0.004			
Implantable cardioverter defibrillator	332 (17)	141 (20)	191 (15)	0.002			
Atrial fibrillation	784 (39)	303 (43)	481 (37)	0.006			
Non-cardiovascular diagnoses							
Hyperkalemia	149 (7)	64 (9)	85 (7)	0.035			
End stage renal disease	136 (7)	77 (11)	59 (5)	< 0.001			
Acute respiratory failure	120 (6)	51(7)	69 (5)	0.076			
Encounter of palliative care	27 (1)	16 (2)	11 (1)	0.008			
Hypertensive chronic kidney disease	130 (7)	64 (10)	66 (5)	< 0.001			
Renal dialysis	86 (4)	44(6)	42 (3)	0.001			
On supplemental oxygen	149 (7)	74 (11)	75 (6)	< 0.001			
Pleural effusion	93 (5)	49 (7)	44 (3)	< 0.001			

Results are shown as mean \pm standard deviation or numbers and percentages.

admission as a categorical variable, and to 0.63 including prior admission as a continuous variable, as illustrated in Figures 3, 4.

When we assessed the length of stay and the recency of prior hospital admission to the index HF admission, we did not find a significant association with 30-day hospital readmission rates or further improvement in the fit of the predictive model (data not shown).

Discussion

We found that hospital admissions in the prior year before an index HF admission predicted 30-day hospital readmission in a real life patient cohort. The rate of 30-day hospital readmission increased with increasing numbers of prior hospital admissions. Data on prior hospital admission yielded an improvement in the accuracy of the 30-day

^{*}Risk of mortality is the likelihood of dying determined at time of admission, scored as one of four categories: minor, moderate, major, and extreme.

**Severity of illness is the extent of physiologic decompensation or organ system loss of function determined at time of admission, scored as one of four categories: minor, moderate, major, and extreme.

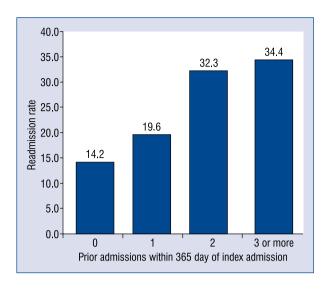


Figure 2. The percentage of patients with 30-day hospital readmission by prior admissions.

hospital readmission predictive model. However, the length of stay or recency of prior admission did not play a role in predicting readmission. Our findings have important clinical implications in identifying a high-risk group for 30-day hospital readmission that could be potentially targeted for interventions to reduce readmissions.

We found a decreased likelihood of 30-day hospital readmission among those with LVEF < 40%, that is consistent with another recent study by Hummel et al. [11]. The possible explanation of this finding is the relatively narrow range of treatment options for HF patients with a preserved ejection fraction (LVEF > 40%) compared to those with a reduced ejection fraction (LVEF < 40%) [12].

Deriving and validating readmission models that are specific to hospital systems is beneficial for hospitals with focused strategies to reduce readmissions [13]. This is especially important since prior studies focused predominantly on Medicare patients that may not reflect local practices with other cohorts.

The discrimination of our readmission model is comparable to previously published 30-day hospital readmission models. Even though there have been previous studies to predict 30-day hospital readmission in HF patients, only one of these studies utilized prior admissions in their model [7].

Table 2. The effect of the number of prior hospital admission as a continuous covariate on 30-day hospital readmission in a multivariate logistic regression model including significant predictive factors.

Effect	Odds ratio est	Р		
	Point estimate	95% Wald confidence limits		
Preserved left ventricular ejection fraction ≥ 40%	1.28	1.01	1.64	0.045
Nonhypertensive	1.40	1.08	1.82	0.011
Number of prior hospital admission in last 365 days	1.31	1.21	1.42	< 0.001

The model is further adjusted for age and race (forced in the model).

Table 3. The effect of prior hospital admission as a categorical covariate on 30-day hospital readmission in a multivariate logistic regression model including significant predictive factors.

Effect	Odds ratio estimates from logistic regression			Р
	Point estimate	95% Wald confidence limits		
Preserved left ventricular ejection fraction ≥ 40%	1.28	1.01	1.64	0.045
Non-hypertensive	1.40	1.08	1.82	0.011
No prior hospital admission in last 365 days	1.00			
One prior hospital admission in last 365 days	1.47	1.08	1.99	0.015
Two or more prior hospital admissions in last 365 days	2.93	2.19	3.92	< 0.0001

The model is further adjusted for age and race (forced in the model).

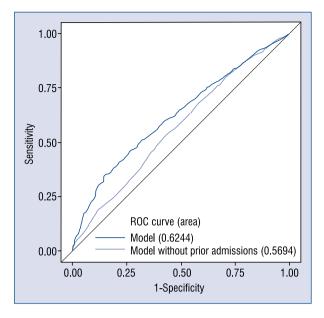


Figure 3. Receiver-operating characteristic (ROC) comparison curves for readmission model to predict 30-day hospital readmissions with and without prior hospital admission modeled as any prior admissions.

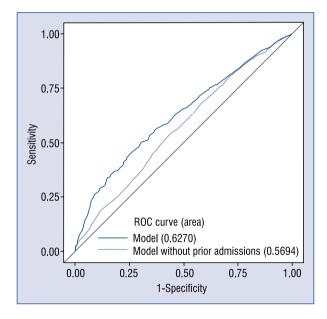


Figure 4. Receiver-operating characteristic (ROC) comparison curves for readmission model using baseline clinical covariates to predict 30-day hospital readmissions with and without prior hospital admission modeled as prior admissions (0, 1, 2 or more).

However, the study by Hummel et al. [7] analyzed only Medicare patients, and therefore the external validity of their findings is limited. In the current

analyses, we have demonstrated the importance of prior admissions to predict 30-day hospital readmission in "all-comers", both Medicare and non-Medicare patients.

Similar to a retrospective analysis by Chin and Goldman [14], our study confirmed that the diagnosis of hypertension was protective of readmission. While their study and ours share similarities and focus on a single hospital system, our analysis encompasses a significantly larger patient cohort.

Another study by Eapen et al. [8] derived and validated a logistic regression model on 10 independent predictors of readmission in a large cohort of patients from the Get with the Guidelines Heart Failure Registry. They found a similar predictive accuracy of their model, with a C-statistics of 0.59. Unlike our study, however, the authors did not include prior admissions in their analysis.

Our study did not address the effect of outpatient follow up on readmission. In a recent report, Casalino et al. [15] found that smaller primary care-based practices had lower rates of potentially preventable hospital admissions compared to larger practices. Outpatient follow-up and physician practice size may have had an effect on the rate of readmission in our population, but our study was not designed to measure this effect. Future studies may benefit from identifying how this factor relates to prior admissions and other health outcomes.

Hummel et al. [7] utilized data from the GAP-HF (Guidelines Applied in Practice-Heart Failure) study and showed that the use of prior admissions improved the discrimination (C-statistics of 0.71) of the 30-day readmission prediction model in Medicare patients with HF. Although the authors were able to demonstrate an improvement in an established model utilizing prior admissions, their study included a patient cohort from an earlier time. Management and treatment of HF has significantly improved since then, limiting the validity of their observations. Our study, however, utilized a more current patient population, which may reflect the advances in management of patients with HF over the last 10 years.

Limitations of the study

Our study has potential limitations. First, the lack of follow-up data for deaths within 30-days could affect the estimates for 30-day hospital readmission after discharge from the hospital. To determine the impact of this bias on our estimates, we performed a sensitivity analysis utilizing the first admission for each patient in the dataset rather than one at random, confirming similar results to

the initial models. In addition, there may have been patients who were readmitted to another hospital system, and they may have been erroneously grouped in the no 30-day readmission group.

Conclusions

Our study shows that hospital admission within the year prior to an index admission for HF significantly increases the risk of 30-day hospital readmission due to any cause. The rate of 30-day hospital readmission increased with a higher number of prior admissions. Our findings may help clinicians identify patients at high-risk for 30-day hospital readmission following an index HF admission to optimize care and enhance follow-up to improve outcomes.

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Conflict of interest: None declared

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Supplementary Table 1. Clinical characteristics of the total patient population, and stratified by 30-day hospital readmission.

Parameters	All patients	Patients with 30-day readmission	Patients without 30-day readmission	Р
Number of patients	1,999	366 (18%)	1,633 (82%)	
Demographics				
Age [years]	68.2 ± 15.6	67.0 ± 16.6	68.5 ± 15.4	0.232
Female gender	824 (41)	163 (45)	661 (40)	0.154
White race	1,485 (74)	259 (71)	1,226 (75)	0.088
Married	920 (46)	157 (43)	763 (47)	0.184
Baseline risk scores				
Risk of mortality* (major or extreme)	950 (48)	196 (54)	754 (46)	0.010
Severity of illness** (major or extreme)	1,366 (68)	273 (75)	1,093 (67)	0.004
Prior hospital admission before index adr	mission			< 0.001
0 prior admissions	1,299 (65)	185 (50)	114 (68)	
1 prior admissions	382 (19)	75 (21)	307 (19)	
2 or more prior admissions	318 (16)	106 (29)	212 (13)	
Characteristics of index admission				
Diagnoses on admission	15.8 ± 6.1	17.3 ± 6.3	15.6 ± 6.0	< 0.001
Admission type emergency	1,720 (86)	333 (91)	1,387 (85)	0.003
Length of stay in days	7.9 ± 15.2	10.1 ± 18.5	7.4 ± 14.3	< 0.001
Disposition to home	1,706 (85)	314 (86)	1,392 (85)	0.788
Left ventricular ejection fraction > 40%	895 (48)	181 (53)	714 (47)	0.046
Co-morbidities at index admission				
Hypertension	714 (36)	109 (30)	605 (37)	0.009
Dyslipidemia	847 (42)	147 (40)	700 (43)	0.344
Ischemic cardiomyopathy	718 (36)	126 (34)	592 (36)	0.510
Left bundle branch block	122 (6)	16 (4)	106 (7)	0.126
Implantable cardioverter defibrillator	332 (17)	69 (19)	263 (16)	0.202
Atrial fibrillation	784 (39)	153 (42)	631 (39)	0.263
Non-cardiovascular diagnoses				
Hyperkalemia	149 (7)	42 (12)	107 (7)	0.001
End stage renal disease	136 (7)	39 (11)	97 (6)	0.001
Hyposmolality and or hyponatremia	145 (7)	36 (10)	109 (7)	0.035
Acute respiratory failure	120 (6)	21 (6)	99 (6)	0.813
Hypertensive chronic kidney disease	130 (7)	39 (11)	91 (6)	< 0.001
Urinary tract Infection	209 (10)	49 (13)	160 (10)	0.043
Depressive disorder	252 (13)	52 (14)	200 (12)	0.307
Family history of diabetes mellitus	217 (11)	53 (15)	164 (10)	0.014
Hypokalemia	247 (12)	49 (13)	198 (12)	0.507
Renal dialysis	86 (4)	28 (8)	58 (4)	< 0.001
On supplemental oxygen	149 (7)	38 (10)	111 (7)	0.018
Pleural effusion	93 (5)	26 (7)	67 (4)	0.014
Acute kidney failure	373 (19)	79 (22)	294 (18)	0.112
Diverticulitis with hemorrhage	182 (9)	42 (11)	140 (9)	0.081
Hypoxemia	220 (11)	53 (14)	167 (10)	0.019

Results are shown as mean ± standard deviation or numbers and percentages.

^{*}Risk of mortality is the likelihood of dying determined at time of admission, scored as one of four categories: minor, moderate, major, and extreme.

**Severity of illness is the extent of physiologic decompensation or organ system loss of function determined at time of admission, scored as one of four categories: minor, moderate, major, and extreme.