

ORIGINAL ARTICLE

Interventions as an alternative to penalties in preventable readmissions

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Received: January 23, 2015

Accepted: March 31, 2015

Online Published: April 28, 2015

DOI: 10.5430/jha.v4n3p70

URL: <http://dx.doi.org/10.5430/jha.v4n3p70>

ABSTRACT

While expenditures in healthcare in the United States are the highest in the world, it is widely known that those resources are not being used efficiently. The government addressed this situation in the Patient Protection and Affordable Care Act, in an attempt to improve quality and affordability of healthcare. In the fiscal year 2013, the Centers for Medicare and Medicaid Services began imposing financial penalties through the Inpatient Prospective Payment System to hospitals that have higher than expected readmission rates for specific diseases. The nature and effects of this new policy have raised several concerns. This article discusses Medicare's hospital readmissions reduction program and presents an alternate policy based on disease-specific interventions to reduce preventable readmissions. Our results show that a policy based on implementing disease-specific interventions, instead of penalties, may save 33.43% of hospitals from being under the penalization level in the first year, while at the same time improving the delivery of care.

Key Words: Disease-specific interventions, Readmissions, Medicare

1. INTRODUCTION

Over the last four years, the United States (US) spent on average 17.74% of its gross domestic product (GDP) on healthcare, the highest in the world. This is more than twice the rate of other high-income countries which was 7.78%.^[1] However, this high expenditure has not translated into a better quality. To illustrate, in 2001 the performance of the US healthcare system was ranked 37th and in 2014 it was classified 46th in efficiency.^[2,3] Furthermore, in 2012 the Institute of Medicine estimated that 30% of the total expenditures in 2009 on healthcare were wasted.^[4] In summary, the US healthcare system continues to face significant challenges in performance, quality and cost.

It has been argued that preventable readmissions are evidence of the deficiency in the quality of care,^[5,6] generating potential harm to patients and unnecessary costs.^[7] Much and varied research has been done to identify the causes of readmissions, their validity, and their interaction when predicting the risk of readmissions.^[8,9] Furthermore, preventable readmissions affect the Medicare-covered population. In fact, Jencks *et al.* in 2009 estimated that between 2003 and 2004, 19.6% of patients were unexpectedly readmitted, representing in 2004 a cost of \$17.4 billion.^[10] The Robert Wood Johnson Foundation in 2013 estimated that readmissions represent \$26 billion for Medicare, of which \$17 billion is estimated as the cost of avoidable readmissions.^[11]

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The US began a reform process promising a better, and less costly, healthcare system.^[12] The Patient Protection and Affordable Care Act (ACA) established that payments to eligible hospitals will be adjusted as an incentive to reduce readmissions. Therefore, the Centers for Medicare and Medicaid Services (CMS) started the hospital readmission reduction program (HRRP) which includes a set of economic penalties through the inpatient prospective payment system (IPPS) to hospitals that show excessive readmissions in certain diseases. However, imposing financial penalties to incentivize hospitals in reducing preventable readmissions has raised concerns such as the appropriateness of the policy or the possible undesired effects to hospitals.

This study explores the feasibility and preliminary effects of a disease-specific intervention as an alternative to HRRP. The goal is to decrease readmissions and reduce cost while directly improving the quality of care.

First, a review of literature is done to prove the plausibility

of the assumption that interventions reduce the amount of preventable readmissions. Secondly, a different review is conducted to describe the available results of interventions in one specific disease. Then, using the data available from CMS, a simulated case is proposed and results from it are calculated. Finally, the simulation results are studied and compared with the available results from the current HRRP policy.

2. BACKGROUND

2.1 Inpatient prospective payment system

The IPPS, introduced in section 1886 (d) of the Social Security Act, is used by Medicare to reimburse hospitals for inpatient care services provided to covered patients. In the IPPS, the reimbursement calculation depends mainly on the diagnosis of the admission (not procedures), represented by the diagnosis related group (DRG) weight. The calculation of this payment is shown in Equation 1.

$$DRG_{weight}(i, j) = (Labor \times WIF(j) + Non Labor) \times DRG_{weight}(i), \forall i \in I, \forall j \in J \quad (1)$$

Where I and J represent the set of providers and hospitals considered in IPPS, respectively. $DRG_{weight}(i)$ is a weight that accounts for the differences among the i diseases in terms of resources and procedures. The $WIF(j)$ term accounts for the socioeconomical differences in each geographic location, and the labor and non-labor wage relates to the different portions of expenses related to the medical service provided. Medicare also adjusts for factors such as longer stays, disproportionate care hospital, indirect medical education, etc. The payment before adjustment is referred here as DRG base payment ($DRG_{base}(i, j), \forall i \in I, \forall j \in J$).

2.2 Calculations for the excess of readmissions

The next element considered in the HRRP is excess of readmissions for the following conditions: acute myocardial infarction (AMI), heart failure (HF) and pneumonia (PN). The excess is calculated using patient-level administrative data for three years. The application of HRRP for FY2013 uses data from FY2009, FY2010 and FY2011. A hierarchical logistic regression is implemented to account for the average effect among hospitals, offering a risk adjustment approach. The expected readmissions measure, the denominator, is obtained by regressing the specific patient-level data using the average intercept while the numerator is obtained using the average intercept and the specific “residual” for each hospital (42 C.F.R. §412.150 - §412.154).

2.3 Hospital readmission reduction program (HRRP)

In the IPPS final rule for FY2013, an adjustment factor (AF) is applied to all reimbursements billed to Medicare from hospitals that present an excess of readmissions for AMI, HF and PN.^[13] In FY2015, total knee arthroplasty (TKA), total hip arthroplasty (THA) and congestive obstructive pulmonary disease (COPD) are included in the calculations of the HRRP. The AF depends on the DRG base payment for each specific disease (AMI, HF and PN in FY2013), the number of cases in the period considered, the payments for all admissions made in the period and the excess of readmissions for AMI, HF and PN (see Equation 2).

This AF affects the total payment for all admissions billed to Medicare through IPPS during the fiscal year. The implementation of the AF considers a ceiling adjustment of 1% for FY2013, which was raised to 2% by FY2014 and 3% for FY2015 (w/o quotations).

From the beginning the methodology, effects and results of HRRP policy attracted criticism. Some of the concerns relate to the inappropriateness of the nature of the incentive,^[14] the impact on the most vulnerable hospitals^[15] or the adjustments of payments applied to all diseases based on a small portion of them.^[16] Also, it is unclear whether the reduction in the payments to hospitals will improve the quality of care.

$$Adj.Factor(j) = 1 - \frac{\sum_{i \in I} \{[Excess(i, j) - 1] \times DRG_{base}(i, j) \times NOC(i, j)\}}{DRG_{all\ admissions}(j)}, \forall j \in J \quad (2)$$

Where $Adj.Factor(j)$ represents the final adjustment applied to reimbursements, $Excess(i, j)$ is the ratio calculated by CMS, $DRG_{(all\ admissions)}(j)$, $\forall j \in J$ represents the payments for all of the admissions for each specific hospital during the period, and $NOC(i, j)$ is the number of cases of each disease by hospital.

2.4 Interventions

Joynt & Jha in 2012 suggest that through holistic approaches, better financial and clinical outcomes can be achieved.^[17] The literature over the last two decades shows examples of improvement in quality of care and reduction in readmissions from interventions.^[18,19] As an illustration, we screened three scientific databases for systematic literature reviews that compile clinical trials of these interventions on patients with HF. This condition was selected since it is linked with the biggest readmissions rate for Medicare patients. Table

1 shows that disease-specific interventions appear to reduce readmissions (281 randomized trials or 64%), thus our analysis builds on this assumption.

Naylor *et al.*, in 2004 concluded that interventions improving the transition of care in elderly patients would bring better clinical and financial outcomes.^[20] Moreover, Hernandez *et al.*, in 2010 found that early follow-up procedures among HF patients lowered their risk of being readmitted.^[21] The documented interventions focus mainly in the discharge process, follow-up process and the transition of care.^[20–25] A review of the literature presenting results of interventions on HF patients is conducted (see Table 2). One of the conclusions from the review is that interventions to reduce readmissions in HF patients do not only improve the desired outcome, but also (in some cases) generate savings. Based on these results, a scenario where an intervention is applied to HF patients is simulated.

Table 1. Literature screening results supporting our assumptions

Authors	Analysis	Timeframe	Subjects	Intervention	No of studies	Conclusion
Holland <i>et al.</i> , 2005 ^[31]	Systematic Literature Review.	Origin to June 1, 2004	Patients with HF.	Multidisciplinary interventions.	74 RCT	Reductions in mortality and admissions. Decreased overall
Jovicic <i>et al.</i> , 2006 ^[32]	Systematic Literature Review.	Origin to Nov. 2005	Patients with HF.	Self-Management Interventions.	6 RT	hospital readmissions and readmissions for heart failure.
Phillips <i>et al.</i> , 2004 ^[33]	Meta-Analysis.	Until Oct. 2003	Patients with CHF.	Comprehensive discharge planning plus post discharge support.	18 RT	Significant reduction in readmissions.
Roccaforte <i>et al.</i> , 2005 ^[34]	Meta-Analysis.	Until 2004	Patients with HF.	Disease Management programs.	33 RT	Reductions in mortality and admissions.

3. METHODS

In the proposed simulated scenario, an intervention is applied to all HF admissions under the IPPS of Medicare. The effects on the AF, as well as the costs, are analyzed to compare the results of implementing this disease-specific intervention with HRRP.

3.1 Simulated intervention

The intervention used in the simulated scenario consists on a single follow-up call for HF patients, made by a registered nurse. The provider checks with the patient or caregiver the adherence to the discharge plan, listens to any change in patient condition or new symptoms, adjusts the medications and suggests visit/s to the hospital as necessary. The

intervention is planned to take one hour (30 min preparation, planning and recording results, and another 30 min of direct communication with the patient). The direct cost of the intervention is based on the time spent by the nurse. The mean annual and hourly wages for a registered nurse is \$67,930 and \$32.66 respectively;^[26] therefore, the cost of the intervention is estimated at \$32.66. The effectiveness of the follow-up call made to HF patients is estimated using the actual reduction results published for similar interventions and included in Table 2. A triangular distribution is fitted to the data compiled from these cited interventions, resulting in a mean effectiveness, *i.e.*, reduction on 30-day preventable readmissions, of 35.8%.

Table 2. Literature summary of interventions on HF patients

Author	Design	Setting	Participants	Intervention	Results	Conclusions
Chaudhry et al., 2010 ^[35]	†	33 cardiology practices around US.	1,653 patients recently hospitalized with HF.	Tele-monitoring.	Readmissions reduced by 3.85%.	Conclusions no non- statistically significant.
Cline et al., 1998 ^[19]	Prospective Randomized trial.	University hospital with a primary catchment area of 250,000 habitants.	190 patients (aged 65-84 years, 52.3% men), hospitalized with HF.	Education on disease, self-management and follow-up and nurse directed outpatient clinic for one year after discharge.	Longer time to readmission (141 (81) vs. 106 (101); $p < .05$). Savings of \$1,300 per patient annually.	Intervention decreased readmissions and costs.
Fonarow et al., 1997 ^[36]	†	†	214 patients with advanced HF.	Comprehensive heart failure management program.	85% reduction in readmissions. Savings of \$9,800 per patient.	Intervention decreased readmissions and admissions for cardiac transplant.
Giordano et al., 2009 ^[37]	Randomized trial.	Cardiovascular rehabilitation departments of "Salvatore Maugeri" Foundation.	460 patients (57 +/- 10 years old) hospitalized with chronic heart failure.	Use of a portable device able to transfer a one lead trace to a cardiologists.	36% decrease in readmissions. Lower cost of readmissions (843 +/- 1,733 vs. 1,298 +/- 2,322).	One year HBT reduced readmissions and cost for CHF patients.
Grafft et al., 2010 ^[22]	Retrospective review of discharges.	Mayo Clinic hospitals in Rochester, MN.	4,989 discharges.	Hospital follow-up appointment.	†	Non appear improvement in readmissions rates.
Hansen et al., 2013 ^[38]	Semi controlled pre-post study.	11 hospitals varying in location, size and academic affiliation.	Target older adults.	Toolkit.	13.6% reduction in readmissions.	Intervention appeared to be associated with a decrease in readmissions rates.
Harrison et al., 2011 ^[23]	Retrospective cohort study.	†	30,272 patients.	Post discharge telephonic follow-up within 14 days after discharge.	23.1% less likelihood to be readmitted in the intervention group.	Intervention is effective at reducing hospital readmissions and, thus, generate potential savings.
Hernandez et al., 2010 ^[21]	Observational analysis.	Network of 225 hospitals.	30,136, 65 years or older patients with HF.	Physician follow-up.	Readmissions in the higher quartile of follow-up 20.9% versus 23.3% in the lower quartile.	Patients with higher physician follow up are less likely to be readmitted.
Jack et al., 2009 ^[39]	Randomized trial.	General medicine service at an urban, academic, safety-net hospital.	749 English speaking hospitalized adults (mean age, 49.9 years).	Nurse based follow-up, med. reconciliation, patient education. Pharmacist telephonic follow-up.	Intervention group had lower rate of hospital utilization (0.314 vs. 0.451)	Intervention reduced hospital utilization within 30 days of discharge.
Krumholz et al., 2002 ^[40]	Prospective Randomized trial.	Yale-New Haven hospital (YNHH).	88 patients (> 50 years old) with HF on admission between 10/1997 and 09/1998.	2 phases: comprehensive evaluation and education. Follow-up sessions.	39% decrease in readmissions. Saving of \$7,515 per patient.	Intervention reduced readmissions and costs for patients with HF.
Naylor et al., 1999 ^[41]	Randomized clinical trial.	Two urban academically affiliate hospitals in Philadelphia, PA.	363 patients, hospitalized between 08/1992 and 03/1996 that had one severe medical and surgical reason for admission.	Advanced nurses deliver a comprehensive discharge planning and home follow-up protocol designed for elders at risk of poor outcomes.	Time to readmission increased in the intervention group. Fewer multiple readmissions in the intervention group (6.2% vs. 14.5% $p < .001$).	Intervention reduced readmissions and increase the time to be readmitted.
Naylor et al., 2004 ^[20]	Randomized clinical trial.	6 Philadelphia academic and community hospitals.	239 patients, 65 years old or older, with HF.	A 3 month APN-directed discharge planning and home follow-up protocol.	Time to readmissions longer in intervention group. Fewer readmissions in the intervention group (104 vs. 142) and lower costs (\$7,636 vs. \$12,481).	Intervention increase the time to readmission, reduce the number of hospitalizations and costs.
Rich et al., 1995 ^[42]	Prospective randomized trial.	Jewish hospital at Washington university medical center.	282 High risk patients, 70 years old or older, hospitalized with HF.	Comprehensive education, special diet, social service consultation, planning for an early discharge, review of medications and an intensive follow-up.	Readmissions was reduced by 56.2%. The cost of care was \$460 less per patient in the intervention group.	Intervention improved quality of life and reduced hospital use and medical costs.
Riegel et al., 2002 ^[43]	Randomized controlled clinical trial.	†	358 patients with CHF.	Telephonic case management.	45.7% lower readmissions at 3 months, 47.8% lower at 6 months. Inpatient heart failure cost 45.5% lower at 6 months.	Intervention reduced readmissions and costs for patients. Results comparable to other pharmaceutical therapies.
Stewart, et al., 1998 ^[44]	Randomized trial.	Tertiary referral hospital that services a largely elderly population.	97 patients with CHF.	Single home visit (by a nurse and pharmacists) to improve medication management, identify clinical deterioration and modify follow-up and care giver vigilance.	Intervention group had lower risk of readmissions (odds ratio 0.4; 95% CI, 0.2-1.1). Intervention group had fewer days of hospitalization (261 vs. 452; $p = .05$).	Among a cohort of high-risk patients with CHF, intervention reduced frequency of unplanned readmissions plus out-of-hospital deaths within 6 months of discharge from the hospital.
Stewart et al., 2002 ^[45]	Prospective evaluation of two randomized studies.	Tertiary institution with a specialist cardiology unit.	297 patients with CHF.	First study: a structured visit by a nurse and a pharmacist. Second study: repeated visits.	Intervention had fewer readmissions (0.17 vs. 0.29 per month, $p < .05$). The median cost of these readmissions was A\$325 versus a A\$660/month.	Intervention benefits in reducing the frequency of unplanned readmissions persist in the long term and are associated with prolongation of survival.

Note. †Not explicitly mentioned in the study

3.2 Data

The data used to study the effect of the simulated intervention comes mainly from public use files (PUF) from repositories available on the CMS website. Specifically, we used the hospital readmissions reduction programs supplemental data file and the Inpatient Medicare Provider Utilization and Payment Data for IPSS FY2013 final rule.^[27] The number of hospitals considered was 3,500.

3.3 Procedure

The data contains the number of cases and excess readmissions for AMI, HF and PN, by provider and the AF for each hospital, considering the adjustment ceiling of 1% for FY2013. Then, it follows that to calculate AF we only need the base payment for each specific condition and the total payments for all admissions. However, in this study there was no access to the total payment for all admissions. In-

stead, we used the DRG, WIF, Labor and non-labor wages to calculate the base payment for each HRRP condition, and having the AF we use both quantities to estimate the total payments for the period. Finally, the excess of readmissions for HF after the intervention is calculated, and the AF is updated.

4. RESULTS

4.1 Base payments

Using the described data, the base payments for the conditions considered by HRRP are calculated for each provider. The labor and non-labor wage for FY2013 were \$3,679.95 and \$1,668.81 respectively. The last two components are the WIF (specific for each hospital) and the DRG weights for each specific diagnosis. Table 3 shows the DRG weights considered in these calculations. The average payment, before the inclusion of the DRG weight, is \$6,431.92.

Table 3. DRG codes and weight for IPSS final rule FY2013

Code	Description	Weight	Average DRG base payment
280	Acute myocardial infarction with multiple comorbidities	1.799	\$ 11,576.81
281	Acute myocardial Infarction with comorbidities	1.096	\$ 7,050.03
282	Acute myocardial Infarction without comorbidities or multiple comorbidities	0.773	\$ 4,975.73
291	Heart failure with multiple comorbidities	1.517	\$ 9,759.80
292	Heart failure with comorbidities	1.003	\$ 6,453.79
293	Heart failure without comorbidities or multiple comorbidities	0.675	\$ 4,342.19
193	Pneumonia with multiple comorbidities	1.489	\$ 9,579.06
194	Pneumonia with comorbidities	0.999	\$ 6,429.35
195	Pneumonia without comorbidities or multiple comorbidities	0.707	\$ 4,552.51

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4.2 DRG base payment for all discharges

After the DRG base payments for each condition and for each hospital is obtained, the DRG base payment for all admissions by each hospital is computed. Results of the DRG payments for all admissions present big differences (see Table 4, showing wide variation among the hospitals serving Medicare populations).

4.3 Results from the simulated scenario

The intervention was applied to the 1,193,210 admissions for HF reported to Medicare through IPSS in FY2013. Results

show that after the intervention, 710 hospitals were freed from penalization, representing a decrease of 33.43% (see Table 5). The average AF also improved from 0.0042 to 0.0039.

Table 4. Descriptive statistics for DRG all admission payments

Measurement	Value
Max	\$ 711,552,145.07
Mean	\$ 1,616,546.71
Min	\$ 70,208,431.47
St Dev.	\$ 72,055,027.96

Figure 1 shows the AF before and after the intervention for 150 providers randomly selected from the 3,500 initially considered. The behavior of the AFs was not homogeneous. Some hospitals experienced high improvement (*i.e.* provider 46), medium (*i.e.* provider 42), or no improvement (*i.e.* provider 63). Additionally, there are hospitals that after im-

plementing the intervention were free from penalizations (*i.e.* provider 49), whereas others improved less and were unable to avoid the penalties (*i.e.* provider 72).

The intervention had a total cost of \$38.9M, while the total amount of penalties was \$253.3M. Comparing these simulated results with the actual HRRP results for the same period, a decrease of \$26.7M is observed (HRRP penalties in FY2013 were \$280M).

Table 5. Hospitals being penalized before and after the intervention

	Hospitals Penalized	% of the total	Average adjustment factor
Before Intervention	2,124	60.69	0.0042
After Intervention	1,414	40.40	0.0039

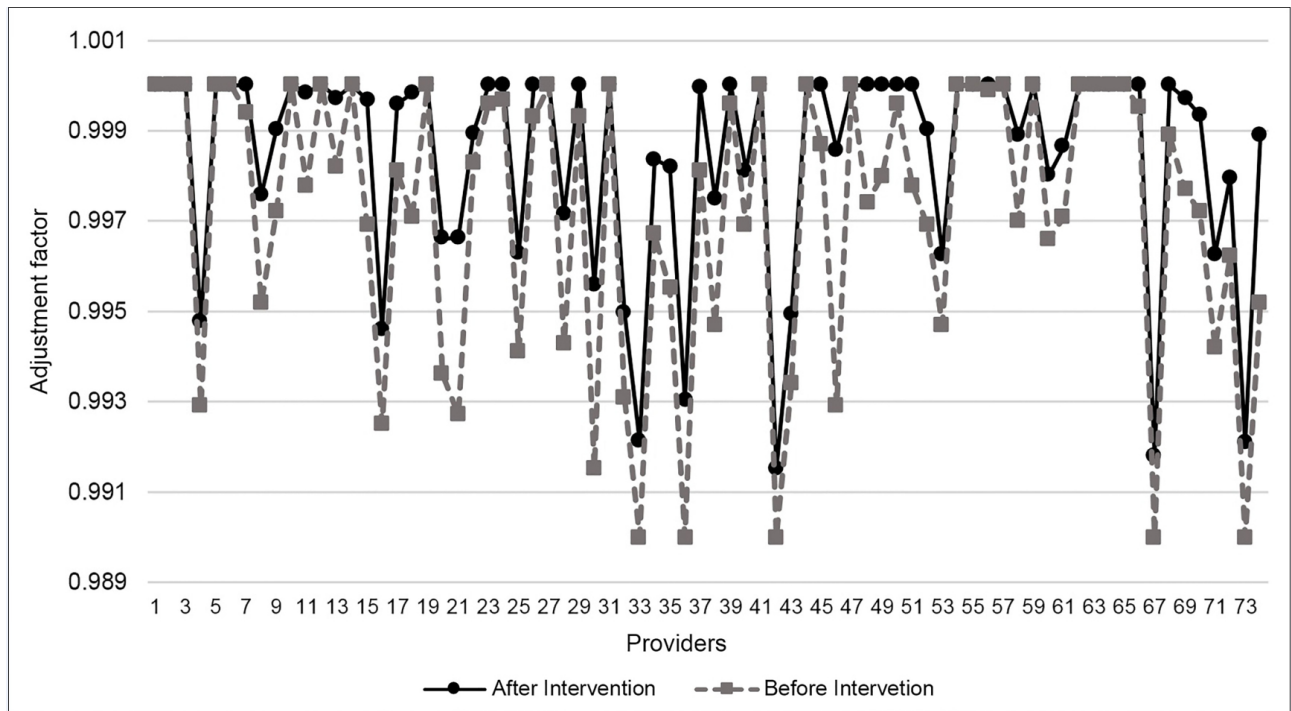


Figure 1. Adjustment ratio after and before the intervention

5. DISCUSSION

A disease-specific intervention approach was presented as an alternative to HRRP, which is known to reduce preventable readmissions as well as to improve the quality of the delivery of care.

HRRP has been in place for three years and during that period the number of hospitals penalized has increased (see Table 6). The differences between FY2013 and FY2014 indicate that in FY2014, 11 more hospitals were penalized, while in FY2015, 413 more hospitals are penalized. These results would suggest that HRRP is not resulting in a decrease of readmission as the number of hospitals being penalized continues to increase. Conversely, a disease-specific intervention would immediately show progress by diminishing the risk of readmission for the patients, which would mean less hospitals being penalized. Results from the simulated scenario show that 710 hospitals are freed from penalties when implementing an intervention. When compared to the

HRRP results for FY2013-FY2014, the simulated intervention drastically outperforms the results of HRRP.

Table 6. History of penalties through the HRRP

Title	FY2013	FY2014	FY2015
n	3,500	3,483	3,476
Cap	1%	2%	3%
Hospitals penalized	2,214	2,225	2,638
Average penalty	0.42%	0.38%	0.62%

Furthermore, a comparison of the AF between the hospitals obtaining “better results” (lower penalty) and hospitals with “worse results” (higher penalty) under HRRP, show that the number of hospitals improving decreased, while the number of hospitals that worsen increased (see Table 7). This represents a contradiction when compared with the mission of CMS which is “better healthcare, better health and lower costs through improvement”.^[28] A disease-specific inter-

vention would ensure an improvement on readmission rates which would lead to, as explained by the current metrics, the number of hospitals being penalized to decrease.

Table 7. Evolution of hospitals’ condition in HRRP

Title	FY2013-FY2014	FY2014-FY2015
Got Worse	1,054	2,024
%	31%	59%
Got Better	1,364	680
%	40%	20%

Meanwhile, a consistent decrease in the excess of readmissions is reported for HF throughout FY2013-FY2015. AMI also shows a decrease in the readmission rate, but just during FY2013-FY2014, while in FY2014-FY2015 there is no improvement. Reductions are found to be inconsistent for

PN, as excess readmissions increased in FY2013-2014, and then decreased (see Table 8). The approach based on disease-specific interventions shows an improvement on preventable readmissions.

Considering the short timeframe that HRRP has been active, results show small and inconsistent improvements in reducing readmissions. Furthermore, it has been said that economic penalties affect more those hospitals that provide care to vulnerable patients and institutions that take the responsibility to teach and train physicians. Results from the simulation show that an approach based on disease-specific interventions would be more appropriate than HRRP because: 1) it outperforms HRRP in reducing the readmission rates; 2) by its very nature improves the quality of the delivery of care; and 3) disease-specific interventions are less costly than the penalties from HRRP.

Table 8. Evolution of hospitals’ condition in HRRP

	AMI			HF			PN		
	2013	2014	2015	2013	2014	2015	2013	2014	2015
Fiscal Year rule	2013	2014	2015	2013	2014	2015	2013	2014	2015
Number of cases	500,931	492,346	505,702	1,193,210	1,161,629	1,154,060	955,611	951,383	971,906
Average (SD)	0.648 (.484)	0.644 (.484)	0.644 (.483)	0.890 (.324)	0.888 (.328)	0.879 (.334)	0.894 (.320)	0.897 (.315)	0.892 (.320)
Change	-	-0.62%	0%	-	-0.22%	-1.01%	-	+0.34%	-0.56%

Additionally, we presented several concerns with the methodology used by HRRP. Stone & Hoffman in 2010 point out that since hospitals bill Medicare for each discharge, there is an incentive in maximizing the discharges.^[14] Moreover, reducing readmissions also reduces the hospital’s revenue, which creates a conflict. A disease-specific intervention not only leads to better quality care but also translates into savings for hospitals. Increased quality of care will also lead to savings for patients as number of hospital readmissions decreases.

Joynt & Jha in 2013 found that the effects of HRRP penalties would be more severe for large hospitals, teaching hospitals and safety net hospitals.^[29] Teaching hospitals represent about 25% of all participating hospitals in the IPPS. Therefore, it can be argued that the penalty approach may negatively impact the quality of medical education in the US. Instead, by applying disease-specific interventions, the quality of care for these patients improves, and avoids the negative financial impact on the hospitals. Furthermore, Berenson *et al.* in 2012 recognize that AMI, HF and PN represent about 12% of Medicare expenditures.^[16] This means that in the 2013 final rule, the 12% of Medicare admissions affected the reimbursement of all the admissions billed to Medicare through the IPPS. Since disease-specific interventions focus

on improvement, say by targeting excessive preventable readmissions, it could eliminate the notion of applying across the board penalties.

Finally, Burgess & Hockenberry in 2013 state that HRRP penalties will worsen the financial situation for those hospitals likely to be affected the most: large hospitals, teaching hospitals and safety net hospitals.^[30] Instead, the authors advise that a policy targeting the causes of the readmissions may produce better results. The implementation of disease-specific interventions has the potential to address preventable readmissions from the mentioned perspective.

Key limitations of this study are: its short timeframe and the absence of patient-level data, which forced the use of aggregate data. Consequently, these results are not yet generalizable. However, the study does suggest (and reinforces) that an approach based on disease-specific interventions should lead to better results, better quality and less cost than HRRP. Another limitation of the this study was that the simulated scenario is applied to all hospitals, neglecting the idea that different hospitals might require different interventions.^[5] However, the authors believe that these initial results encourage further work in this direction.

Future work, in addition to addressing the limitations stated

above, could also include other disease-specific interventions, considering the unique reality, characteristics and needs of specific hospitals (or cluster of hospitals). Granted this requires access to more granular, hospital/patient specific, data. Additionally, the implementation of disease-specific interventions should ideally be as patient centered as possible. It is very likely that to properly design, model and analyze

these efforts researchers will require the development and implementation of probabilistic models or decision support systems, which include patient specific data. Consequently, having access to hospital and patient-level data will enable more realistic modeling and simulation strategies that would lead to stronger and more robust implementable conclusions.

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