TECHNICAL APPENDIX

Methods Notes

Socioeconomic Status Measure

Evidence suggests that hospital readmission rates vary based on the racial, ethnic, and socioeconomic status mix of patients served by the hospital.¹⁻⁴ Linked 2009 MedPAR and Medicare Beneficiary Annual Summary files were used to determine the proportion of hospital patients who self-identified as Black or Hispanic (separately). We also determined the proportion of hospital admissions with Medicaid as the primary payer from the AHA data. The average socioeconomic status indicator for zip codes within the hospital service area was derived from 2006-2010 American Community Survey data using a composite measure of median household income, median value of housing units, proportion of households with interest, dividend, or rental income, proportion of adult residents completing high school, proportion of adult residents completing college, and proportion of employed residents with management, professional, and related occupations.^{5, 6}

Matching approach

We used an optimal nonbipartite matching approach⁷⁻⁹ to carry out a comparison of HRRP penalties between well-staffed hospitals versus otherwise similar hospitals with poor staffing levels. To create matched pairs, we categorized hospitals based on five quintiles of our nurse staffing variable.^{7, 10, 11} We estimated a propensity score for each observation using an ordinal logit model.^{7, 8} Subsequently, we generated a distance matrix where the distance between any two hospitals in the same nurse staffing quintile was set to infinity to preclude matching

within quintile. The distance formula was adjusted by dividing by the square of the difference in quintile so that the distance between two hospitals became smaller as the staffing quintile difference increased.⁷ Thus, we created the optimal set of one-to-one matches that minimized the overall distance between pairs of hospitals while giving preference to pairs that differed most in nurse staffing. We allowed the matching algorithm to ignore up to 5% of outliers hospitals to avoid matching substantially dissimilar hospitals, i.e., hospitals that lack overlap in their distribution of covariates. This would require us to rely on extrapolation for inferences about the average treatment effect of staffing on the penalties.^{12, 13} Following matching, we designated the better staffed hospital in each matched pair as having "high" staffing and the more poorly staffed hospital as having "low" staffing. We used Lu, et al.'s⁸ *nbpMatching* package in R.

Balance diagnostics and Cross-match test

Before conducting any analysis or looking at any data related to the outcome, we examined balance diagnostics to ensure that our matching procedure resulted in hospital pairings that were equivalent in all respects except for staffing. First, we calculated absolute standardized differences between the high and low staffing groups to evaluate covariate balance after matching.¹⁴ We also used the cross-match test to assess multivariate balance.¹⁵ The cross-match test evaluates how well our matching compares to a randomized control trial in terms of balancing covariates; i.e., was covariate balance across groups with matching roughly similar or greater as that expected when each hospital in the pair is randomly assigned to either the high or low staffing group. The cross-match procedure uses nonbipartite matching to pair hospitals based on the covariates but ignoring treatment (in this case, staffing) category. The test then counts the cross-matches or the instances when a matched pair comprised a treated (high staffing) hospital and a control (low staffing) hospital based on the original matching procedures. The idea is that

if treated hospitals are rarely paired with control hospitals using covariates alone, the covariate distributions of the treated and control hospitals must differ. The likelihood of a cross-match in a randomized control trial is one half, so any greater proportion of cross-matches signifies greater than expected balance. Our cross-match statistic of 0.60 (p = 1.00) indicates good covariate balance and suggests that if two hospitals with the same propensity score were chosen at random, one would have come from the high staffing group and one from the low staffing group 60% of the time — better than the 50% probability expected in a randomized control trial.

Alternative specifications

We repeated analyses specifying the outcome to contrast hospitals that were fully penalized with those penalized less than the full 1%. We also conducted analyses without matching on the skilled nursing facility indicator. We repeated our analysis using FY2012 HRRP data and the September release of FY 2013 data to ensure that our findings were not the result of an isolated occurrence. We also used logistic regression without matching but adjusting for covariates. Each of these model results are presented in Appendix Exhibit C.

APPENDIX EXHIBIT A



HRRP readmissions penalty

SOURCE. Authors' analysis of data from 2,826 hospitals. **NOTES**. Each dot represents one hospital. The x-axis represents the percentage payment reduction where 0 represents no penalty and 1 represents the full 1% penalty.

APPENDIX EXHIBIT B

Characteristics of Hospitals Before and After Matching

		Pre-Matching				Post-Matching			Standardized Difference in Means
Variable	Initial Sample (N=2,976)		Analytic Sample (N=2,826)		Low Staffing (N=1,413)		High Staffing (N=1,413)		(Low vs. High)
	Number or mean	% or SD	Number or mean	% or SD	Number or mean	% or SD	Number or mean	% or SD	%
Registered nurse hours per adjusted day	6.4	2.5	6.5	2.3	5.1	1.5	8.0	2.0	164.1
Covariate balancing score	_	_	_	_	7,170	13,320	7,170	13,320	0
Covariates									
Number of beds Resident/fellow	227	204	224	198	219	188	229	207	5.3
physicians per bed Operating profit	0.06	0.19	0.05	0.16	0.05	0.14	0.06	0.17	5.6
margin Percentage Black	-0.02	0.23	-0.02	0.24	-0.02	0.15	-0.02	0.30	0.9
patients Percentage Hispanic	11.6	15.7	11.6	15.8	11.6	16.0	11.7	15.7	0.9
patients Percentage Medicaid	2.3	5.3	2.4	5.4	2.4	5.7	2.4	5.1	0.5
mix Socioeconomic status	19.1	12.6	17.7	10.4	17.3	10.6	18.0	10.1	6.3
indicator Herfindahl-Hirschman	-0.7	2.5	-0.6	2.5	-0.6	2.6	-0.6	2.4	1.0
Index Technology level–no.	772	1,554	766	1,554	802	1,654	729	1,447	4.7
(%) High	1.040	250/	1.025	260/	408	250/	527	270/	3.5
Low	1,049	55%	1,023	50% 64%	498 915	55%	527 886	57% 63%	
Ownership-no. (%)	1,927	0570	1,001	0170	715	0570	000	0370	0.4
For-profit	578	19%	575	20%	289	20%	286	20%	
Not-for-profit Geographic location– no. (%)	2,398	81%	2,251	80%	1,124	80%	1,127	80%	2.3
Urban	2,655	89%	2,548	90%	1,280	91%	1,268	90%	
Rural Skilled nursing facility–no. (%)	321	11%	278	10%	133	9%	145	10%	4.5
Yes	485	16%	360	13%	167	12%	193	14%	
No	2.491	84%	2.466	87%	1.246	88%	1.220	86%	

SOURCE. Authors' analysis of data from all 2,976 hospitals with data and analytic sample of 2,826 hospitals. **NOTES**. Standardized differences represent differences in means in units of standard deviation between the high and low staffing group for each covariate expressed as a percentage.

APPENDIX EXHIBIT C

Effect of Registered Nurse Staffing on Odds of FY 2013 Hospital Readmissions Reduction Program Penalty

		Adjusted Through Matching		Adjusted Through Matchin and Regression		
	No.	Odds Ratio (95% CI)	P Value	Odds Ratio (95% CI)	P Value	
High vs. low registered nurse staffing	2826	0.75 (0.63-0.89)	< 0.001	0.75 (0.63-0.88)	0.001	
Sensitivity Analyses						
1) Full 1% penalty vs. no penalty or < 1% penalty	2826	0.62 (0.48-0.81)	< 0.001	0.59 (0.44-0.77)	< 0.001	
2) Excluding skilled nursing facility variable	2826	0.79 (0.67-0.93)	0.005	0.78 (0.66-0.92)	0.004	
3) FY 2013 (September 2012 revision)	2826	0.75 (0.63-0.88)	< 0.001	0.74 (0.63-0.88)	0.001	
4) FY 2012	2824	0.76 (0.64-0.90)	0.001	0.74 (0.62-0.87)	0.001	
5) Logistic regression without matching using all hospitals with data	2976	0.95 (0.92-0.98)	< 0.001	0.89 (0.86-0.93)	< 0.001	
6) Logistic regression without matching using matched hospital sample	2826	0.93 (0.89-0.96)	< 0.001	0.90 (0.86-0.93)	< 0.001	

SOURCE. Author's analysis of data from 2,826 hospitals. **NOTES**. Analyses adjusted through matching are McNemar's test for matched, paired data evaluating the odds of readmissions penalty based on high versus low nurse staffing. The analyses adjusted through matching and regression represent conditional logistic models that reintroduce the covariates that were used for matching as covariates in the estimation of the effect of high versus low staffing on the odds of readmissions penalty. The sensitivity analyses follow the same format except use alternative specifications of the outcome (#1), matching without skilled nursing facility indicator covariate (#2), or different versions of the data (#3 & #4). The results for analysis #5 represent unpaired logistic regression model estimates of the effect of staffing (measured as registered nurse hours per patient day) on the odds of readmission penalty in the sample of all 2,976 hospitals with outcome and covariate data; analysis #6 is the same as #5 but is carried out using the 2,826 hospitals that were in the matched sample.

APPENDIX EXHIBIT D





Effect of Unmeasured Confounder on Odds of Being Not Penalized

SOURCE. Author's analysis of results from data on 2,826 hospitals. **NOTES**. Sensitivity analysis based on how unobserved confounders play a role in the odds of a hospital being not penalized and the odds of the hospital having a high staffing level. The x-axis is how much the odds of the hospital being not penalized increases for a high level of the unmeasured confounder and the y-axis measures how much the odds of the hospital having the high level of staffing increases for a high level of the unmeasured confounder. Points between the two curves correspond to effects by unmeasured confounders that result in a p-value of less than 0.05, meaning that there is strong evidence that high staffing reduces penalization. Any points outside the two curves correspond to effects by unmeasured of greater than 0.05.

Analysis of patient reported outcomes

APPENDIX EXHIBIT E

Difference in overall patient satisfaction and reports of receiving information about what to do during recovery at home between matched-pairs of high and low staffed hospitals								
	Low Staffing (N=1,413)	High Staffing (N=1,413)	Average difference between matched- pairs of hospitals	Standardized Difference in Means				
Percentage of patients who gave a rating of 9 or 10 (high)	65%	67%	2%***	28%				
Percentage of patients who would definitely recommend the hospital	67%	70%	3%***	28%				
Percentage of patients reporting, YES, they were given information about what to do during their recovery at home	81%	82%	1%***	22%				

SOURCE. Author's analysis of matched pairs from data on 2,826 hospitals. **NOTES**. ***p < 0.001. Standardized differences represent differences in means in units of standard deviation between the high and low staffing group for each covariate expressed as a percentage.

After evaluating whether the likelihood of readmission penalties differed for well-staffed and poorly staffed hospitals, we evaluated whether there was evidence that patient evaluations of care processes that have been suggested to be associated with readmissions and impacted by nurse staffing levels differed in our two groups of hospitals. We used linked data from the Hospital Consumer Assessment of Healthcare Providers and Systems (HCAHPS) survey data from the April 2009 – March 2010 reporting period with our matched set of hospitals. The HCAHPS survey is a national, standardized, publicly reported survey of patients' perspectives of hospital care that are reported on the Hospital Compare database. The data are aggregated and risk-adjusted before release and are reported publicly as a set of ten measures. An overall HCAHPS score comprises one domain of measurement for determining incentive payments under CMS's Value-Based Purchasing Program.

Our primary goal was to evaluate whether a higher proportion of patients in better staffed hospitals reported being given information about what to do during recovery at home compared to patients in matched poorly staffed hospitals. We also examined overall patient satisfaction using the two global measures from the HCAHPS data: 1) the percentage of patients who gave the hospital a rating of 9 or 10 out of 10 (high), and 2) the percentage of patients who would definitely recommend the hospital to friends and family. We have shown in prior work using earlier HCAHPS data that higher proportions of patients in better staffed hospitals respond positively to these compared to poorly staffed hospitals.^{16, 17} We assessed these global outcomes to confirm that the relationship is consistent in this sample of hospitals and because prior research has shown that they are associated with readmissions.¹⁸ After merging these data, we evaluated matched-pair t-tests to determine the average difference in percentages of patients reporting each measure between well-staffed hospitals and their less well-staffed counterparts. The results (appendix exhibit E) indicated small but significant differences between the well-staffed and poorly staffed hospitals pairs for each of the HCAHPS outcomes.

SUPPLEMENTAL EXHIBIT 3

Distribution of Standardized Balancing Score by High and Low Staffed Hospital Groups After Matching



SOURCE. Authors' analysis of data from 2,826 hospitals. **NOTES**. The figure demonstrates the outcome of matching in terms of achieving similarity across the balancing score for the low staffing and high staffing hospital groups.

SUPPLEMENTAL EXHIBIT 4

Distribution of Registered Nurse Hours Per Patient Day by High and Low Staffed Hospital Groups After Matching



SOURCE. Authors' analysis of data from 2,826 hospitals. **NOTES**. The figure demonstrates the outcome of matching in terms of achieving differences in registered nurse staffing levels between the high and low nurse staffing groups.

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***The following include full model results for primary analyses (results presented in the text of the paper) as well as alternative model specifications (results in Appendix Exhibit C.) We show both McNemar's and bivariate clogit models for the matching only models

****MCNEMARS AND CONDITIONAL LOGISTIC REGRESSION WITH MATCHED PAIRS**** **MCNEMARS WITH MATCHED PAIRS****

Cases		Controls Exposed	Unexposed	 	Total
	Exposed Unexposed	728	247 109	 	975 438
	Total	1057	356		1413

McNemar's chi2(1) = 11.67 Prob > chi2 = 0.0006 Exact McNemar significance probability = 0.0007

	Cases Controls	.6900212	[95	% CI]					
	difference ratio rel. diff.	0580326 .9224219 2303371	0 .8 3	918927 806556 768991	0241 .9661 0837	.724 .691 751			
	odds ratio	.7507599	.6	339595	.8881	562	(exact)		
******** ****CLOG *******	************ ************* IT VERSION **********	********** ********** OF PRIMAF *********	********* ********* Y ANALYS ********	******** ******** IS** *******	* * * * * * * * * * * * * * * * *	*******	* * * *		
Conditio	nal (fixed-	effects)	logistic	regress	sion	Number LR chi2 Prob >	of obs 2(1) chi2	= =	2826 11.71 0.0006
Log like	lihood = -9	973.56028				Pseudo	R2	=	0.0060
high	Odds Rat	io Std.	Err.	Z	P> z	[95% (CI]		
penalty	.75075	.063 .063	2071	-3.40	0.001	.63655	576 .885	545	08

CONDITIONAL LOGISTIC REGRESSION ADJ. WITH MATCHING & REGRESSION ADJUSTMENT Conditional (fixed-effects) logistic regression Number of obs = 2826 LR chi2(13) = 55.60Prob > chi2 = 0.0000Log likelihood = -951.61811Pseudo R2 = 0.0284_____ high | Odds Ratio Std. Err. z P>|z| [95% CI] penalty | .7453978 .0648972 -3.37 0.001 .6284623 .8840909 1.022075.0062923.550.0001.0098171.0344827.73e5.48e3.540.00071767.928.32e bdtot | hightech | 7.73e resprbed | 1.37e 1.52e 3.56 0.000 4.96e 3.78e ownership | 8.96e-08 4.10e-07 -3.55 0.000 1.14e-11 .0007048 margin | 186.883 280.6232 3.48 0.000 9.849332 3545.95 3.44 0.001 105.9615 2.41e urb2 | 50486.89 158840.7 per medcaid | 1.90e-25 3.11e-24 -3.48 0.000 2.30e-39 1.57e-11 blackp | .0000229 .0000695 -3.52 0.000 5.92e-08 .0088236 3.48 0.001 90917.95 5.99e 3.54 0.000 3.07582 50.09253 2.33e hispanicp | 1.76e sesi | 12.41272 8.835677 -3.50 0.000 5.01e-21 1.86e-06 snf | 9.65e-14 8.26e-13 -3.62 0.000 .9987179 .9996188 hhi | .9991683 .0002298

********Full 1% Penalty Vs. No Penalty Or < 1 % Penalty *** ****MCNEMARS AND CONDITIONAL LOGISTIC REGRESSION WITH MATCHED PAIRS**** ****MCNEMARS WITH MATCHED PAIRS--Full 1% Vs. No Penalty Or < 1 % Penalty | Controls Cases | Exposed Unexposed | Total _____+
 Exposed |
 12
 87
 |

 Unexposed |
 140
 1174
 |
 99 1314 _____+ Total | 152 1261 | 1413 McNemar's chi2(1) = 12.37 Prob > chi2 = 0.0004 Exact McNemar significance probability = 0.0005 Proportion with factor Cases .0700637 .1075725 [95% CI] Controls _____ _____ difference -.0375088-.0590235-.0159942ratio.6513158.5119717.8285853rel. diff. -.0420301-.065935-.0181253 odds ratio .6214286 .4699631 .8178539 (exact) **CLOGIT VERSION OF PRIMARY ANALYSIS*Full 1% Vs. No Penalty Or < 1 % Penalty ******* Conditional (fixed-effects) logistic regression Number of obs = 2826 LR chi2(1) = 12.49Prob > chi2 = 0.0004Log likelihood = -973.17226Pseudo R2 = 0.0064_____ high | Odds Ratio Std. Err. z P>|z| [95% CI] _____ penalty | .6214286 .0848361 -3.48 0.000 .47554 .8120736 _____

**Full 1% Vs	$\Delta U \cup U \supset I \subseteq N \cap $									
FULL 10 VS. NO FENALLY OF < 1 7 FENALLY SAMAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA										
Conditional (fixed-effects) logistic regression Log likelihood = -949.97286 Number of obs= 2826 LR chi2(13) = 58.89 Prob > chi2 = 0.0000 Pseudo R2 = 0.0301										
high	Odds Ratio	Std. Err.	Z	P> z	[95% CI]					
<pre>penalty bdtot hightech resprbed ownership margin urb2 per_medcaid blackp hispanicp sesi snf hhi </pre>	.5854989 1.022796 1.75e 5.20e 5.33e-08 218.2741 70704.35 2.80e-26 .000015 5.56e 13.44559 3.59e-14 .9991367	.082786 .0064183 1.27e 5.88e 2.49e-07 334.1061 226717.7 4.67e-25 .0000465 4.27e 9.755734 3.13e-13 .0002341	-3.79 3.59 3.58 3.61 -3.59 3.52 3.48 -3.53 -3.59 3.52 3.58 -3.55 -3.69	0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000	.4437837 1.010294 124470.4 1.23e 5.69e-12 10.86654 131.8309 1.83e-40 3.49e-08 163415.6 3.243192 1.35e-21 .9986779	.7724686 1.035453 2.47e 2.20e .0004987 4384.427 3.79e 4.30e-12 .0064868 1.89e 55.7426 9.57e-07 .9995958				

*MCNEMARS AND CONDITIONAL LOGISTIC REGRESSION WITH MATCHED PAIRS***NOSNF ***MCNEMARS WITH MATCHED PAIRS*******NO SNF | Controls | Exposed Unexposed | Total Cases Exposed | 716 259 | 975 328 Unexposed | 110 | 438 Total | 1044 369 | 1413 McNemar's chi2(1) = 8.11 Prob > chi2 = 0.0044 Exact McNemar significance probability = 0.0050 Proportion with factor Cases .6900212 Controls .7388535 [95% CI] _____ _____ difference -.0488323 -.08305 -.0146145 .933908 .8909704 .978915 ratio rel. diff. -.1869919 -.3271972 -.0467865 odds ratio .7896341 .6683083 .9322065 (exact) ****CLOGIT VERSION-**NO SNF** Conditional (fixed-effects) logistic regression Number of obs = 2826 LR chi2(1) = 8.13Prob > chi2 = 0.0044= 0.0042Log likelihood = -975.35221Pseudo R2 _____ high | Odds Ratio Std. Err. z P>|z| [95% CI] penalty | .7896341 .0656384 -2.84 0.004 .6709184 .9293561 _____

CONDITIONAL LOGISTIC REGRESSION ADJ. WITH MATCHING (no SNF) **AND REGRESSION (SNF included as covariate) Conditional (fixed-effects) logistic regression Number of obs= 2826 LR chi2(13) = 115.28Prob > chi2 = 0.0000Pseudo R2 = 0.0589

Log likelihood = -921.77763

						_
high	Odds Ratio	Std. Err.	Z	P> z	[95% CI]	
penalty bdtot hightech resprbed ownership margin urb2 per_medcaid blackp hispanicp sesi snf hhi		.0671761 .0040029 8.85e 1.10e .0000342 32.58354 2292.571 1.80e-22 .1348526 2.17e 5.616536 .0403107 .0002888	-3.15 2.27 2.14 2.13 -2.13 2.13 2.09 -2.09 -2.09 -2.10 2.01 2.14 -9.03 -2.19	0.002 0.024 0.033 0.033 0.033 0.033 0.036 0.037 0.036 0.045 0.032 0.000 0.029	.6352026 1.001211 3.827843 12.44244 9.40e-11 1.279088 1.52104 1.36e-45 .025137 2.058128 1.173789 .2456047 .9988018	.8998612 1.016902 3.53e 4.33e .3906624 389.8642 348560.2 .0367049 .8795565 1.00e 35.51781 .4052764 .9999339

****CONDITIONAL LOGISTIC REGRESSION ADJUSTED WITH MATCHING (no SNF) **AND REGRESSION ADJUSTMENT (NO SNF) *****NO SNF MATCHING OR REGRESSION**

Conditional (fixed-effects) logistic regression Number of obs =2826 LR chi2(12) =21.75 Prob > chi2 =0.0405 Log likelihood = -968.54419Pseudo R2 =0.0111

high	Odds Ratio	Std. Err.	Z	P> z	[95% CI]	
penalty bdtot hightech resprbed ownership margin urb2 per_medcaid blackp hispanicp sesi hhi	.7791063 1.008341 5503745 2.16e .0000101 18.56472 494.352 4.22e-23 .2094934 9.20e 5.968826 .9993997	.0668009 .0039536 4.15e 3.21e .0000566 26.83052 1541.977 1.07e-21 .1870341 1.37e 5.143728 .000286	-2.91 2.12 2.06 2.07 -2.06 2.02 1.99 -2.04 -1.75 2.00 2.07 -2.10	0.004 0.034 0.040 0.039 0.040 0.043 0.047 0.042 0.042 0.080 0.046 0.038 0.036	.6585887 1.000622 2.080879 4.776689 1.74e-10 1.092688 1.093902 1.29e-44 .0364103 1.733918 1.102434 .9988392	.9216777 1.01612 1.46e 9.77e .5869334 315.414 223405.7 .1382124 1.20536 4.88e 32.31655 .9999605

*****MCNEMARS AND CONDITIONAL LOGISTIC REGRESSION WITH MATCHED PAIRS**** ***** SEPTEMBER DATA *****MCNEMARS WITH MATCHED PAIRS****** SEPTEMBER DATA | Controls | Exposed Unexposed | Cases Total _____ Exposed | 730 246 | 976 Unexposed | 329 108 | 437 _____+ Total | 1059 354 | 1413 McNemar's chi2(1) = 11.98 Prob > chi2 = 0.0005 Exact McNemar significance probability = 0.0006 Proportion with factor .6907289 Cases [95% CI] .7494692 Controls _____ _____ difference -.0587403 -.092568 -.0249125 allference-.0387403-.092368-.0249123ratio.9216242.8799887.9652295rel. diff.-.2344633-.381972-.0869545 odds ratio .7477204 .6312614 .8847296 (exact) ****CLOGIT VERSION***FY2013 - SEPTEMBER DATA RELEASE Conditional (fixed-effects) logistic regression Number of obs=2826 LR chi2(1) =12.02 Prob > chi2 =0.0005 Log likelihood = -973.40555Pseudo R2 =0.0061 _____ high | Odds Ratio Std. Err. z P>|z| [95% CI] penalty | .7477204 .0630242 -3.45 0.001 .633859 .8820348 _____

penalty .	7418896 .0					
bdtot 1 hightech 7 resprbed 1 ownership 9 margin 18 urb2 50 per_medcaid 1 blackp hispanicp 2 sesi 12 snf 9	.022064 .0 .62e 5. .34e 1. .03e-08 4. 86.3128 27 0174.83 15 .96e-25 3. .000023 .0 .33e 1. 2.39519 8. .81e-14 8.	0062907 3 .40e 3 .49e 3 .13e-07 -3 79.7168 3 57830.1 3 .20e-24 -3 0000699 -3 .75e 3 .821571 3 .39e-13 -3	3.42 3.55 3.54 3.56 3.54 3.54 3.54 3.54 3.54 3.54 3.54 3.54 3.54 3.54 3.48 3.51 3.54 3.54 3.54 3.54 3.54 3.54 3.54 3.54	0.001 0.000 0.000 0.000 0.000 0.001 0.001 0.001 0.000 0.001 0.000	.6253325 1.009809 70966.65 4.88e 1.15e-11 9.824468 105.4248 2.39e-39 5.97e-08 90932.95 3.072265 5.11e-21	.8801721 1.034469 8.18e 3.69e .0007092 3533.265 2.39e 1.61e-11 .0088767 5.96e 50.00892 1.88e-06

*****MCNEMARS AND CONDITIONAL LOGISTIC REGRESSION WITH MATCHED PAIRS****FY2012 *****MCNEMARS WITH MATCHED PAIRS******FY2012 | Controls | Exposed Unexposed | Cases Total _____+ Exposed | 738 241 | 979 319 114 | Unexposed | 433 _____+ 1057 355 | Total | 1412 McNemar's chi2(1) = 10.86 Prob > chi2 = 0.0010 Exact McNemar significance probability = 0.0011 Proportion with factor .6933428 Cases .7485836 Controls [95% CI] _____ _____ difference -.0552408 -.0886703 -.0218113 .8849245 .9694138 ratio .9262062 rel. diff. -.2197183 -.3640107 -.0754259 odds ratio .7554859 .6364405 .8958611 (exact) ****CLOGIT VERSION***FY2012 Conditional (fixed-effects) logistic regression Number of obs=2824 LR chi2(1) =10.90 Prob > chi2 =0.0010 Log likelihood = -973.27397Pseudo R2 =0.0056 _____ z P>|z| [95% CI] high | Odds Ratio Std. Err. penalty | .7554859 .0644787 -3.29 0.001 .6391143 .8930467 _____

CONDITIONAL LOGISTIC REGRESSION ADJ. WITH MATCHING & REGRESSION *FY2012 ***** Conditional (fixed-effects) logistic regression Number of obs=2824

Log likelihood = -942.07054

LR chi2(13)	=73.31
Prob > chi2	=0.0000
Pseudo R2	=0.0375

high	Odds Ratio	Std. Err.	Z	₽> z	[95% CI]	
penalty bdtot hightech resprbed ownership margin urb2 per_medcaid blackp hispanicp sesi snf hhi	<pre></pre>	.0651205 .0067896 4.94e 2.09e 1.94e-09 11046.81 1.38e 1.19e-32 9.66e-07 1.28e 22.17025 1.16e-17 .000247	$\begin{array}{c} -3.47 \\ 4.46 \\ 4.47 \\ 4.46 \\ -4.45 \\ 4.51 \\ 4.43 \\ -4.40 \\ -4.43 \\ 4.40 \\ 4.47 \\ -4.45 \\ -4.45 \\ -4.45 \end{array}$	0.001 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000	.6182327 1.016608 2.05e 1.09e 2.87e-14 133.9513 4836.205 1.13e-48 3.61e-10 3.26e 6.672563 1.65e-26 .9984158	.874784 1.043223 2.04e 2.79e 5.56e-06 247444.5 3.33e 4.14e-19 .0002235 6.23e 129.0179 9.50e-11 .9993841

UNMATCHED LOGISTIC REGRESSION EVALUATING EFFECT OF HRS/PATIENT DAY ON PENALTY **USING 2976 HOSPITALS THAT HAD OUTCOME AND COVARIATE DATA** Logistic regression Number of obs = 2976LR chi2(1) = 11.15Prob > chi2 = 0.0008 = 0.0032 Log likelihood = -1762.5555Pseudo R2 _____ penalty | Odds Ratio Std. Err. z P>|z| [95% CI] _____+ hrs_day | .9467159 .0155177 -3.34 0.001 .916785 .9776239 _cons | 3.63677 .4157734 11.29 0.000 2.906714 4.550188 _____ Number of obs = 2976 Logistic regression LR chi2(13) = 204.82Prob > chi2 = 0.0000 Pseudo R2 = 0.0579 Log likelihood = -1665.719_____ penalty | Odds Ratio Std. Err. z P>|z| [95% CI] ______ hrs day | .8997855 .0183749 -5.17 0.000 .8644826 .93653 bdtot | 1.000952 .0003329 2.86 0.004 1.0003 1.001605 hightech | .7409915 .0839291 -2.65 0.008 .5934731 .9251782 resprbed | 3.836227 1.494707 3.45 0.001 1.787521 8.232989 ownership | .5886687 .0719417 -4.34 0.000 .4632814 .7479922 margin | 1.156952 .2181799 0.77 0.439 .7994553 1.674312 urb2 | .93687 .138997 -0.44 0.660 .7004746 1.253044 medcaid | .6962138 .2814595 -0.90 0.370 .3152289 1.537656 blackp | 21.75037 8.975031 7.46 0.000 9.687875 48.83203 per_medcaid | hispanicp | 25.48487 26.21186 3.15 0.002 3.394685 191.3223 sesi | 1.01698 .0198812 0.86 0.389 .9787508 1.056703 .9999795 .0000275 -0.75 0.456 .9999257 1.000033 hhi | snf | .7298916 .0908025 -2.53 0.011 .5719582 .9314348 cons | 5.652269 1.324586 7.39 0.000 3.57065 8.947432 _____

UNMATCHED LOGISTIC REGRESSION EVALUATING EFFECT OF HRS/PATIENT DAY **ON PENALTY USING 2826 MATCHED HOSPITALS ONLY Logistic regression Number of obs = 2826 LR chi2(1) = 18.62 = 0.0000 Prob > chi2 Log likelihood = -1668.9366Pseudo R2 = 0.0055_____ penalty | Odds Ratio Std. Err. z P>|z| [95% CI] _____+ hrs day | .9257766 .0165337 -4.32 0.000 .8939318 .9587558 _cons | 4.264816 .5413265 11.43 0.000 3.325513 5.469429 ____ Number of obs = 2826 Logistic regression LR chi2(13) = 206.45 Prob > chi2 = 0.0000 Prob > chi2 Log likelihood = -1575.0227Pseudo R2 =0.0615 _____ penalty | Odds Ratio Std. Err. z P>|z| [95% CI] ______ .8963059 hrs day | .018683 -5.25 0.000 .8604257 .9336822 bdtot | 1.000664 .0003424 1.94 0.052 .9999932 1.001335 hightech | .7659572 .0876939 -2.33 0.020 .6119994 .9586454 resprbed | 7.873335 4.007368 4.05 0.000 2.90345 21.35026 .5840313 .0717028 -4.38 0.000 .4591265 .7429165 1.21777 .2286369 1.05 0.294 .8428528 1.759457 ownership | margin | urb2 | .8221062 .1329059 -1.21 0.226 .5988504 1.128593 er_medcaid |.7721201.3945263-0.510.613.28362912.101934blackp |25.8354111.220917.490.00011.0285660.52179hispanicp |17.7142718.072652.820.0052.398286130.8415 per medcaid | sesi | 1.02578 .020529 1.27 0.203 .9863228 1.066816 hhi | .9999803 .0000282 -0.70 0.485 .9999251 1.000036 .7918039 .1044381 -1.77 0.077 .6114276 1.025393 snf | cons | 6.571581 1.595571 7.75 0.000 4.083187 10.57646 _____