

Uncovering value in hospital patient surveys

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Challenges to the healthcare business

Providers can't charge higher rates. Payers can't increase premiums. [...] To uncover value and grow business, stakeholders must start to chip away at the cost of care.

- Mitchell Morris, MD, Optum

Main stakeholders:

- insurance companies
- hospitals

Research proposal

Project outline: We link patient survey data to hospital performance metrics to identify means of reducing costs (by improving the quality of healthcare).

Data source: The Centers for Medicare & Medicaid Services (CMS) Medicare provider data (<https://data.cms.gov/provider-data/>). In particular, the datasets *Patient survey (HCAHPS) - Hospital*, *Unplanned Hospital Visits -Hospital*, and *Hospital-Acquired Condition Reduction Program*.

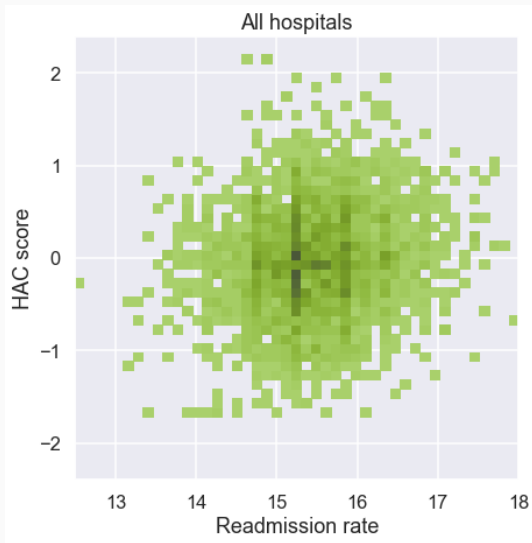
Response variables

- (1) Rate of readmission within 30 days of being discharged from a hospital stay.
- (2) Total Hospital-Acquired Condition (HAC) Score (signed standard deviations from mean).

Hazards to stakeholders:

- Increased costs/claims payments.
- Damage to reputation.
- Reduced funding.
- Greater likelihood of malpractice lawsuit.

Little relationship between targets



Predictors

Amalgamated at the hospital level from patient surveys, they are the linear scores attributed to

- Cleanliness
- Nurse communication
- Doctor communication
- Staff responsiveness
- Communication about medicine
- Discharge information
- Care transition
- Overall hospital rating
- Quietness
- Recommend hospital

Readmission rate: best subsets regression

OLS Regression Results

| | | | |
|-------------------|------------------|---------------------|----------|
| Dep. Variable: | y | R-squared: | 0.091 |
| Model: | OLS | Adj. R-squared: | 0.088 |
| Method: | Least Squares | F-statistic: | 33.81 |
| Date: | Sun, 25 Jul 2021 | Prob (F-statistic): | 2.58e-63 |
| Time: | 13:37:10 | Log-Likelihood: | -3610.9 |
| No. Observations: | 3384 | AIC: | 7244. |
| Df Residuals: | 3373 | BIC: | 7311. |
| Df Model: | 10 | | |
| Covariance Type: | nonrobust | | |

| | coef | std err | t | P> t | [0.025 | 0.975] |
|-------|---------|---------|--------|-------|--------|--------|
| const | 20.6760 | 0.617 | 33.484 | 0.000 | 19.465 | 21.887 |
| x1 | -0.0150 | 0.005 | -3.141 | 0.002 | -0.024 | -0.006 |
| x2 | -0.0090 | 0.012 | -0.734 | 0.463 | -0.033 | 0.015 |
| x3 | -0.0164 | 0.009 | -1.743 | 0.081 | -0.035 | 0.002 |
| x4 | -0.0126 | 0.006 | -2.181 | 0.029 | -0.024 | -0.001 |
| x5 | 0.0206 | 0.005 | 3.818 | 0.000 | 0.010 | 0.031 |
| x6 | -0.0469 | 0.005 | -8.998 | 0.000 | -0.057 | -0.037 |
| x7 | 0.0407 | 0.011 | 3.763 | 0.000 | 0.019 | 0.062 |
| x8 | 0.0392 | 0.015 | 2.539 | 0.011 | 0.009 | 0.069 |
| x9 | -0.0079 | 0.003 | -2.309 | 0.021 | -0.015 | -0.001 |
| x10 | -0.0470 | 0.009 | -5.009 | 0.000 | -0.065 | -0.029 |

| | | | |
|----------------|---------|-------------------|----------|
| Omnibus: | 537.283 | Durbin-Watson: | 1.803 |
| Prob(Omnibus): | 0.000 | Jarque-Bera (JB): | 2526.924 |
| Skew: | 0.688 | Prob(JB): | 0.00 |
| Kurtosis: | 7.004 | Cond. No. | 1.39e+04 |

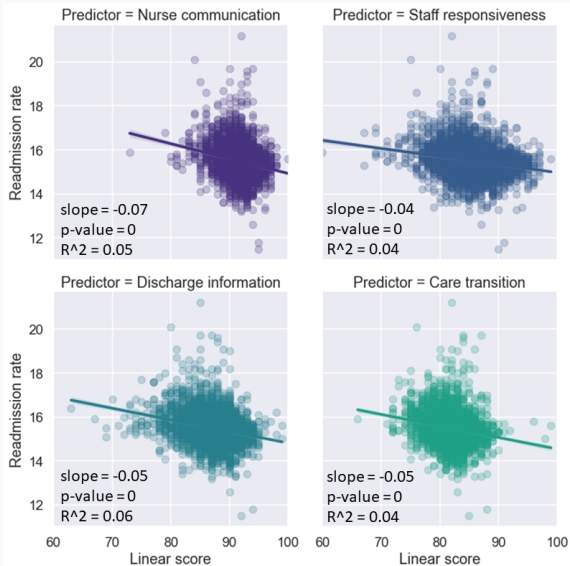
Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.39e+04. This might indicate that there are strong multicollinearity or other numerical problems.

Readmission rate: correlations

| | |
|------------------------------|-----------|
| Discharge information | -0.262240 |
| Recommend hospital | -0.232840 |
| Nurse communication | -0.226153 |
| Overall hospital rating | -0.225432 |
| Staff responsiveness | -0.212882 |
| Care transition | -0.203613 |
| Doctor communication | -0.197653 |
| Cleanliness | -0.193883 |
| Communication about medicine | -0.167898 |
| Quietness | -0.157066 |

Readmission rate: concentrate on these predictors



Total HAC score: correlations

| | |
|------------------------------|-----------|
| Staff responsiveness | -0.227254 |
| Cleanliness | -0.211575 |
| Quietness | -0.202352 |
| Communication about medicine | -0.191847 |
| Care transition | -0.180225 |
| Nurse communication | -0.179704 |
| Overall hospital rating | -0.173021 |
| Doctor communication | -0.165452 |
| Discharge information | -0.143240 |
| Recommend hospital | -0.138713 |

Comparing strongest correlations for the two targets

| Readmission rate | Total HAC score |
|-------------------------|------------------------------|
| Discharge information | Staff responsiveness |
| Recommend hospital | Cleanliness |
| Nurse communication | Quietness |
| Overall hospital rating | Communication about medicine |

The intersection is empty.

Conclusions and recommendations

- The regressions were statistically significant.
- The effect sizes were small.
- Results were qualitatively interesting.

Relatively high readmission rate: try to improve discharge information and nurse communication.

In the data examined, each one point increase in the discharge information score corresponds to a 0.05% decrease in the 30 day readmission rate.

Relatively high total HAC score: try to improve staff responsiveness and cleanliness.

In the data examined, each one point increase in the staff responsiveness score corresponds to a 0.03 s.d. decrease in the total HAC score.

Future work

Improve

- Resolve issue of mismatched collection period dates.
- Better understand the patient survey process and timeline.
- Adapt regression methods to account for multicollinearity.
- Refine approach to address non-constant variance.

Extend

- Segment the data prior to analysis.
- Consider other response variables, such as from the *Complications and Deaths* dataset.

Replace

- Patient level data that would allow for a machine learning approach to predicting which patients are at high risk of readmission.

Thank you.

Any questions?