

# Predictive Analytics to Drive Quality and Efficiency Gains Under Bundled Payment Models

*Sixth National Bundled Payment Summit*

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# Growing Body of Experience

## BPCI Model 2 – Novato Community Hospital

- Started in January 2014 with 30-day LEJR episodes
  - Approximately 100 Medicare cases per year
  - 3 community SNFs
  - Upside only gainsharing with Orthopedic Surgeons

## CJR – 11 Hospitals, spanning two MSAs

- A range of hospital sizes, physician relationships, and post acute partner environments
  - Evolving strategy for supporting best practices

# Challenges and Opportunities

- Independent orthopedists, many in solo private practice
- Episodic cost variability across physicians with individualized protocols
- Variable utilization of post-acute services and providers
- Bundled payments can be a tool to improve quality through MD engagement and financial alignment

# Managing Insurance Risk

- Over a given population, episodic payments must be greater than the combined costs of outcomes
  - Reduce the incidence of high cost outcomes
  - Reduce the median episode costs
- Decreasing the variability of episodic costs has a greater financial return than reducing the costs of the median episode (80/20 rule)
- Quality is measured as a rate of catastrophic events

# Variation Reduction Drives Predictability

- Difficult to make significant traction due to differences in MD training, comfort, expectations, etc.
- All variation is not created equal, but care pathways are complex and it's difficult to identify causal factors
- Perception that standardization devalues the art and clinical judgment of medicine

# Foundation for Long-term Success

*How do I most rapidly and sustainably reduce episodic variation?*

*Solution:*

- **Identify meaningful sources of variation**
- **Deliver targeted actions**
- **Continuous monitoring**

# METHODS

# Approach Overview

- Suite of predictive models that identify specific person-place-time processes and actors associated with variability in outcomes
- *Direct process redesign efforts for greatest impact*

## Optimize SNF Utilization

Identify predictive patient attributes (like unmarried and female) and direct preoperative planning and patient education

## Optimize Hospital Length of Stay

Predictions based on patient attributes (like comorbidity status, and age) to direct inpatient interventions and process improvements

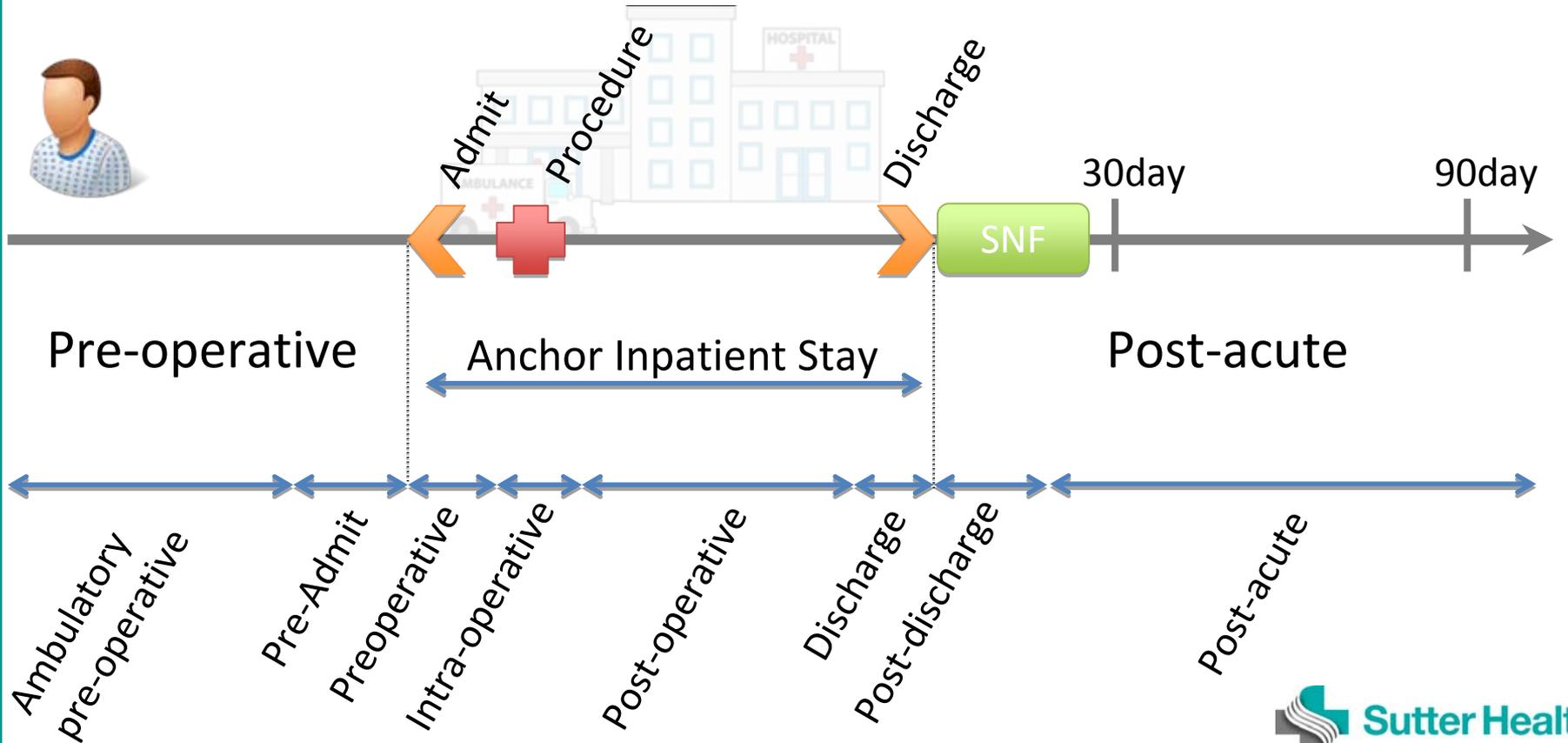
## Minimize Hospital Readmissions

Identify and prioritize modifiable drivers of readmission in the first 30 days post acute, and over the remaining 31 to 90 days

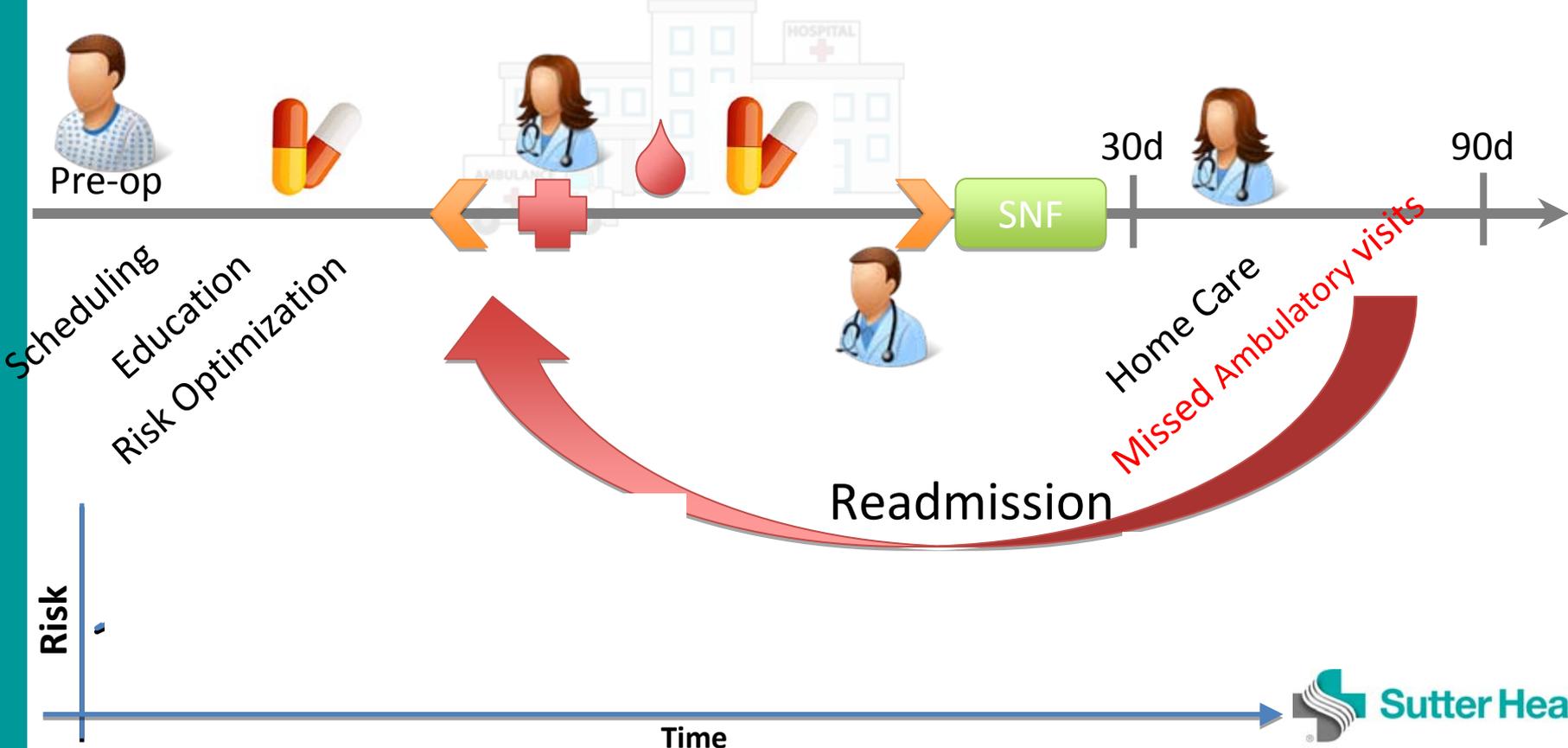
# Analytic Methods

- Analyze a set of episodes (by DRG)
- Model explicit and surrogate features
- Identify meaningful, phase-specific features and interactions within and among phases
- Focus first on explained variation and dominant person, place, and time sources
- Elastic Net Regression – extremely fast penalized regression
  - Regularization path is computed scanning between the lasso and elasticnet
  - Shrinks the model, identifying important predictors, among numerous redundant predictors
  - Better predictive power than ordinary regression
- Models trained with cross validation, performance reported on 25% reserved as a test set

# Phases of an Episode



# Dissecting a Patient's Episode Timeline



# IDENTIFY SOURCES OF VARIATION

# Identified 10's of Risk Factors per Hospital

DISEASE_CATEGORY ANALYSIS	h1p				h2p				h3p				h4p				h5p							
	disp_snr	los	readmit_30d_any	readmit_90d_any																				
(Intercept)	-7.108	2.749	-3.570	-3.641	-4.124	1.396	-3.570	-3.641	-1.729	1.963	-3.570	-3.641	-0.470	2.524	-3.570	-3.641	-7.812	2.568	-3.570	-3.641	-1.303	-4.066	-3.570	-3.641
Age (years)	0.064				0.022	0.011			0.005				0.061	0.003			0.045				0.116			
Alcohol user																								
Asian Race																								
Black Race																								
BMI < 16.5																								
BMI at least 30 and less than 40																								
BMI at least 40																								
Count of diagnoses not present at admit	0.093	0.488			0.038	0.175			0.213				0.671								0.671			0.671
Count of diagnoses present at admit	0.066				0.025	0.048			0.038				0.023								0.003			-0.038
Count of prior ED and admissions	0.052				0.178	0.11			0.033				0.023								0.176			0.176
Discharge disposition against medical advice																								
Discharge disposition of routine home																								
Discharge disposition to home with home health																								-0.029
Discharge disposition to SNF																					0.165			0.165
DRG is 469																					-0.016			0.036
DRG is with fracture																								-0.016
Emergency admit	1.057	2.930																			1.392			1.392
Female	0.322																							-0.032
Female & not married	0.642																							
History of anxiety																								
History of depression																								
History of diabetes																								
Illicit drug user																								
Inpat_abnorm_lab_Calcium_Low			0.037																					0.037
Inpat_abnorm_lab_Chloride_High	0.132																							
Inpat_abnorm_lab_Creatinine_High			0.071																					0.071
Inpat_abnorm_lab_Glucose_High	0.006																							
Inpat_abnorm_lab_Hematocrit_Low				0.155																				
Inpat_abnorm_lab_Hemoglobin_Low	0.309			0.163																				
Inpat_abnorm_lab_Platelet_Count_Low																								
Inpat_abnorm_lab_POC_Glucose_High	0.049																							0.088
Inpat_abnorm_lab_RBC_Low																								0.128
Inpat_abnorm_lab_Sodium_Low	0.198																							-0.074
Inpat_abnorm_lab_UA_Blood_Abnormal				0.088																				0.222
Inpat_abnorm_lab_UA_Epithelial_Cells_Abnormal				0.220																				0.354
Inpat_abnorm_lab_UA_Ketones_Abnormal																								0.598
Inpat_abnorm_lab_UA_PH_High																								0.079
Inpat_abnorm_lab_Urea_Nitrogen_Bun_High	0.02																							0.094
Inpat_abnorm_lab_WBC_high																								0.541
Inpat_abnorm_lab_A1CPERCENT_High																								0.236
Inpat_abnorm_lab_ALBUMIN_Low																								0.102
Inpat_abnorm_lab_ANIONGAP_Low																								0.503
Inpat_abnorm_lab_CHLORIDE_Low																								0.132
Inpat_abnorm_lab_DIFF_LYMPHOCYTES_Low	0.104																							0.511
Inpat_abnorm_lab_DIFF_MONOCYTES_High	0.103																							0.132
Inpat_abnorm_lab_DIFF_NEUTROPHILS_High																								0.511
Inpat_abnorm_lab_GFFRESTAFRICANAMERICAN_Low																								0.104
Inpat_abnorm_lab_GFFRESTOTHER_Low	0.054																							0.103
Inpat_abnorm_lab_INR_High																								0.103
Inpat_abnorm_lab_MCHC_Low																								0.103
Inpat_abnorm_lab_OSMOLALITYCALC_SERUM_Low																								0.103
Inpat_abnorm_lab_POTASSIUM_Low																								0.103
Inpat_abnorm_lab_QBOYTIME_PATIENT_High																								0.103



*Is our case mix unfavorable?*

*Which operational issues to prioritize?*

# Example - Optimizing SNF Utilization

Dashboard	Model Results	Care Redesign	Cost Savings
HOSPITAL	Hospital Site A		
DISEASE	lejr ▼		
ANALYSIS	disp_snf ▼		
ANALYSIS DATE	3 selected ▼		
Cost Driver	Cost Driver Coefficient	Population Mean	
(Intercept)	-8.470	1.000	
Age (years)	0.061	69.394	
Alcohol user	-0.434	0.750	
BMI at least 40	0.343	0.039	
Count of diagnoses not present at admit	0.010	0.475	
Count of diagnoses present at admit	0.008	4.897	
Count of prior ED and admissions	0.200	0.054	
Female	0.010	0.569	
Female & not married	1.072	0.245	

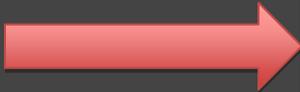
The odds ratio of unmarried females being discharged to SNF is 2.9

Note: Sample data indicative of our findings

# Example - Optimizing SNF Utilization

Dashboard	Model Results	Care Redesign	Cost Savings
HOSPITAL			
Hospital Site A			
DISEASE			
lejr ▾			
ANALYSIS			
disp_snf ▾			
ANALYSIS DATE			
3 selected ▾			
Cost Driver	▲	Cost Driver Coefficient ▾	Population Mean ▾
(Intercept)		-8.470	1.000
Age (years)		0.061	69.394
Alcohol user			0.750
BMI at least 40			0.039
Count of diagnoses not present at admit			0.475
Length of stay (days)			2.681
Payer is Commercial			0.412
Payer is Medicare			0.588
phys_px_id_2556		-0.226	0.216
phys_px_id_3113		0.579	0.387
phys_px_id_3265		0.091	0.074

Orthopedic Surgeon #3113 is associated with higher risk of SNF utilization, controlled for other factors, with odds ratio 1.8



Note: Sample data indicative of our findings

# TARGETED ACTIONS

# Targeted Actions Mitigate Variation

*Optimize SNF utilization with two actions:*

- 1) Better support of unmarried females
  - Pre-operative discharge planning
  - Pre-operative home health visits
  - “Pre-hab” clinic
- 2) Engage Physician 3113 with data and training on standardized practices
  - Support his/her adoption of standardized practices

# Project Financial Impact of these Two Actions

Model Results	Care Redesign	Cost Savings
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## Aggregate Predication Rate

Analysis ▲	Historical ⚡	Y1 ⚡	Y2 ⚡	Y3_5 ⚡
disp_snf	0.159	0.112	0.107	0.105

## Episode Cost Savings Per Case

Analysis ▲	Projected Cost ⚡	Y1 ⚡	Y2 ⚡	Y3_5 ⚡
disp_snf	\$(1,587)	\$470	\$512	\$536

# CONTINUOUS MONITORING

# Continuous Monitoring

- Success requires on-going reduction of variation (in practice and utilization)
- We must know the effectiveness (e.g. ROI) of each program that we launch
- Actor performance varies over time

# Monitoring in Practice

Model Results

Care Redesign

Cost Savings

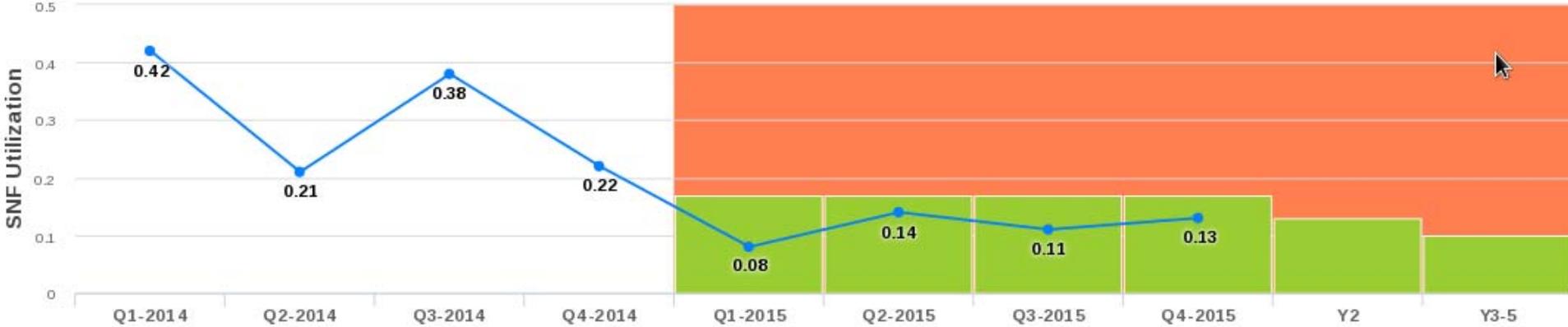
Gainshare & Margin

Success Metrics

More ▾

Disposition to SNF

Disease : *lej*r Hospital : Hospital A



Note: Sample data indicative of our findings

# Closing Remarks

- Bundled payment programs present opportunities, and some unique challenges that call for analytics
- Our analytic framework empowers hospital operators to
  - *Identify sources of variation*
  - *Deliver targeted actions*
  - *Continuously monitor*
- We look to target specific patients for support programs soon
- We will report results from a pilot later this year