Using Machine Learning to Understand Predictors of Frequent Surveillance Testing in Colorectal Cancer Patients

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Background

- Post-treatment surveillance is crucial for colorectal cancer (CRC) patients to detect recurrence early.
- Guidelines recommend regular carcinoembryonic antigen (CEA) testing every three to six months for five years post-treatment.
- However, patients often do not receive recommended surveillance testing. Understanding real-world surveillance patterns may help enhance post-treatment care for CRC survivors
- At a given surveillance visit, if providers could identify patients who have a low likelihood of returning for the next visit on time, they could potentially intervene. This presents an opportunity to improve adherence to guidelines and enhance patient care.

Objectives

- This study aims to identify key patient and physician characteristics that predict the likelihood of a patient returning for frequent CEA testing in current practice.
- By accurately predicting the probability of a patient returning for the next visit on time, healthcare providers can intervene on patients who have a low likelihood of returning for their next surveillance visit.

Methods

Southern California EHR database.

Patients: 2,489 adult patients diagnosed with AJCC stage I-III CRC between 2008 and 2013.

Outcome: At each visit, we defined the binary outcome of whether the next surveillance CEA test occurred within three months. We treated multiple testing visits during surveillance as a recurrent event.

Candidate Features and Feature Engineering:

- Baseline and time-varying patient variables, such as prior and current CEA values
- Created time-varying covariates for time since last surveillance CEA visit, time since surveillance start, and time since last surveillance imaging
- Derived biomarker-related summary measures, such as trend in biomarker levels over time

patient and physician characteristics.

training set.

key predictors.



- Model Fitting: Utilized Logistic Regression with Lasso Penalty. Conducted variable selection among 252
- Internal Validation: Split-sample validation approach, dividing data into 75% training and 25% validation sets.
- **Cross-validation:** Determined optimal regularization parameters using 10-fold cross-validation within the
- Model Evaluation: ROC curves and calibration plots.
- Interpretability: Conducted feature importance analyses to enhance model interpretability and identify

Results



- The Lasso model has good discriminatory power in predicting patients' likelihood of undergoing CEA testing within the next three months. Importantly, the Lasso model demonstrates good calibration, indicating close alignment between predicted probabilities and observed outcomes.
- Patients were less likely to undergo testing within three months if it had been longer since the start of surveillance and if they were diagnosed with Stage 1 or 2 disease. They were more likely to undergo testing if it had been longer since last imaging and they had a high prior CEA value.

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Lasso

Penalized Regression: Variable Selection (Average Beta Coefficients)



Conclusions

- This research sheds light on current surveillance patterns and may be helpful in designing interventions that deliver appropriate, value-based post-treatment care for CRC survivors
- By accurately predicting the probability of a patient returning for the next visit on time, this model could also be used in practice by providers to intervene on patients who have a low likelihood of returning for their next surveillance visit.

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• **Timing between between surveillance and CEA testing, CEA levels, and stage**

QUICK FACTS

Study Title: Using Machine Learning to Understand Predictors of Frequent Surveillance Testing in Colorectal Cancer Patients

- Study Type: Retrospective Cohort Study
- Data Source: Kaiser Permanente Southern California EHR Database
- Patient Population: Adult patients diagnosed with AJCC stage I-III CRC between 2008 and 2013
- Outcome of Interest: Biomarker testing during surveillance, treated as a recurrent event
- Analytical Methods: Logistic Regression with Lasso Penalty and XGBoost models
- Key Findings: Lasso model moderately predicts likelihood of frequent CEA testing; XGBoost model demonstrates strong accuracy
- Implications: Insights can guide post-treatment care strategies for CRC survivors

Calibration Curve (XGBoost) 1.0 - XGBoost 0.4 0.6 Mean predicted probability 0.6 0.8

1.0

XGBoost

Variable Importance Plot (Information Gain)

	0	0.2	0.4	0.6
Days from last cea visit				
AJCC T stage: T2				
Perineural Invasion: Present				
Language: English				
Days since last imaging to cea				
Days from surveillance to first visit				
More than 6 positive lymph nodes				
MSA: Baldwin Park Medical Center				
BMI				
Specialty: Surgery, General				
Age				
CEA slope				
CEA previous visit				
Perineural Invasion: Not Present				

• **XGBoost** demonstrates strong accuracy in identifying patients who are likely to undergo CEA testing within the next three months, with an exceptional AUC. The calibration plot reveals a deviation from the expected calibration, suggesting potential challenges with the model's calibration. Future work will consider recalibration techniques to improve the alignment between