Telehealth Utility Model: a Tool for Improving Patient Care in Rheumatology

Duke AI HEALTH

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Data Science Fellowship Program

Background

- The COVID-19 pandemic catalyzed the rapid adoption of telehealth in rheumatology practices worldwide.
- Our group developed a novel scoring system called the Encounter Appropriateness Score for You (EASY), in which providers document their perception of the appropriateness of telehealth or in-person care after every encounter. EASY score is defined:
 - 1. Either in-person or telehealth acceptable
 - 2. In-person preferred
 - 3. Telehealth preferred
- An analysis of over 12,000 EASY scores in our practice demonstrated that approximately 30% of all follow-up visits could be done by telehealth.

Goal

Expand access to telehealth care by developing a
predictive model that could identify future visits
appropriate for telehealth rather than in-person care via
the EASY score.

Data

 Analyzed return visits to Duke Rheumatology with a documented EASY score.

Methods

Model:

- Predicts telehealth appropriateness via EASY score defined as a binary variable:
 - (1) Telehealth
 - (0) In-Person
- Developed with 10-fold cross-validated logistic regression on a randomized 70/30 train/test partition.
- Covariates include (1) Patient sociodemographic characteristics, (2) Patient Clinical Characteristics, (3) Encounter Characteristics and (4) Provider Characteristics.

Pilot Implementation:

- Presented 4 providers with future visits from 3/1/22 7/31/22 that the model recommended a change in visit modality (n=201).
- Half of the presented recommendations are real suggestions to change from the model, the other half are artificially recommended changes.
- The providers could accept or reject the proposed change in encounter type.
- If the provider accepted the change, the patient is given an opportunity to accept or reject the proposed change of their upcoming visit modality.



Results - Model Performance

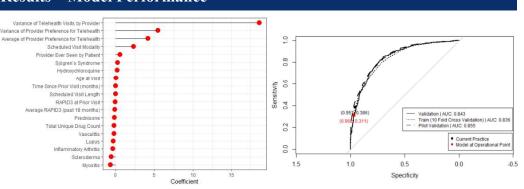


Figure 1: Dot chart of the coefficients of the 19 statistically significant covariates in the telehealth utility model. Sorted from largest to smallest coefficient

Figure 2: ROC curves for the Train, Validation and Pilot Validation

Data Cohort	Telehealth Predicted	AUC	Average Precision	Recall	Accuracy
Train 1/1/2021 – 12/31/2021	0.150	0.836 (0.828, 0.844)	0.711 (0.691, 0.723)	0.373 (0.358, 0.390)	0.770 (0.762, 0.778)
Validation 1/1/2022 - 1/31/2022	0.117	0.843 (0.819, 0.866)	0.703 (0.653, 0.750)	0.311 (0.263, 0.358)	0.769 (0.746, 0.794)
Pilot Validation 3/1/2022 - 7/31/2022.	0.069	0.855 (0.843, 0.866)	0.705 (0.673, 0.722)	0.250 (0.227, 0.276)	0.807 (0.796, 0.817)

Table 1: Performance metrics for the Train, Validation and Pilot Validation data cohorts.

Results – Pilot Implementation

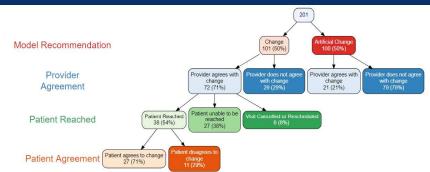


Figure 3: Flowchart of provider and patient agreement during the stages of pilot implementation.

- Percent of provider agreement among the true changes from the model 72/101 (71%) was higher than the rate of agreement with the randomly selected artificial changes 21/100 (21%)
- Unweighted Cohen's Kappa = 0.50, p<0.001 provides evidence that there is moderate agreement between the provider and
 the model on whether the encounter modality should be changed
- Of the 72 visits in which the provider accepted a true change, the patient was reached by clinic staff in 38 visits (54%). Of
 those, 27/38 (71%) patients agreed to the recommended change, where 26 of the 27 visits changed from In-Person to
 Telehealth.

DEI Discussion

 Gender and Insurance variables were statistically significant predictors during model development. However, model performance was not impacted when these variables were removed from the model.

Conclusion

- Provider level aggregate covariates are the most impactful in increasing the probability of a telehealth prediction.
- Consistent performance of the model across all of the data cohorts supports the goal for the model to maintain current performance for future predictions on the target population.
- Rates of patient agreement with the true model recommendations shows that patients are willing to change their visit modality as a result of the model's suggestion.

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