Readmission Prediction of Diabetic Patients

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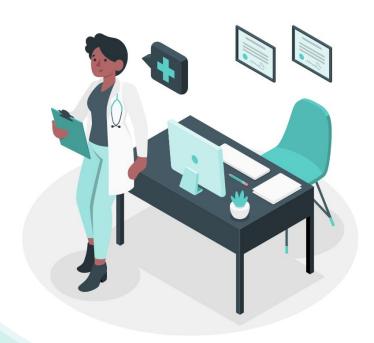


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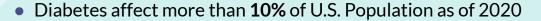
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INTRODUCTION



- Type II diabetes account for over 90%
- 7.8 million hospital discharges of adults in U.S. had diabetes in their diagnosis in 2016
- 16 million emergency visits in 2016 had diabetes as a list of diagnosis
 - Over 10% of total emergency visits in U.S. in 2016
- Diabetes directly and indirectly contributed to over 350,000 death in 2017



Source - CDC, diabetesresearch.org, acepnow.com



DATA



130 Hospitals



Over 50 Features



102,000 Patients



Predictant: No readmission, <30 days, and >30 days

METHODS



Data Acquisition

Data Prep

EDA

Downloaded from Kaggle based on VCU research Cleaned missing data

Visualization

Recategorized features

Statistical Test

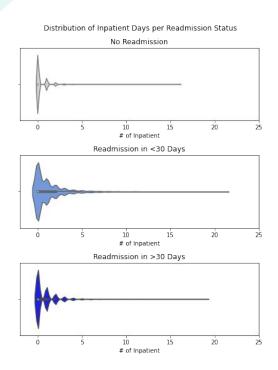
Models: Decision Tree, Random Forest

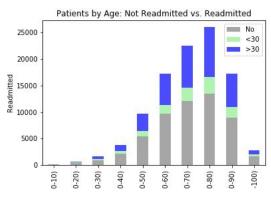
Metric: Recall, "micro" averaging

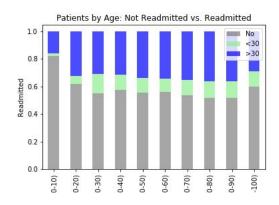
Hypertune: Grid Search Cross Validation

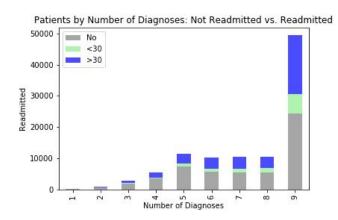
Score: Random Forest: 0.942

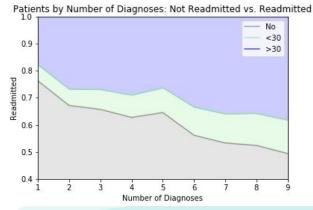
MODELING - Features



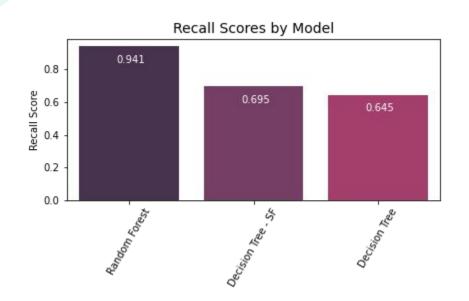


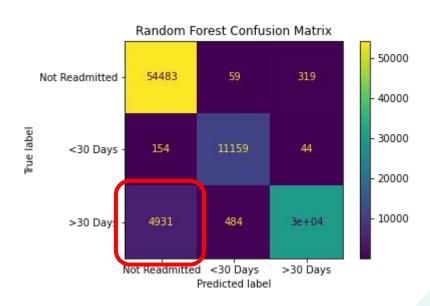






MODELING - Performance





Conclusions



- Health Care Providers can plan future treatments based on model output.
- Patients with expected readmission greater than 30 days can plan for future check-ups.
- Patients with expected readmission within 30 days can receive more focused treatment.
- Health Care Providers can plan for potential overcrowding and resource allocation.

Future Steps

- Check for new treatment method for additional features.
- Predict on recent data to see change in treatment or patient trends.
- Check for **diabetes types** to check for difference in feature importance.



Thank You

Questions?

Project Repo

