

Readmission Prediction of Diabetic Patients

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INTRODUCTION

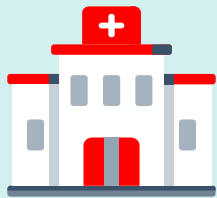


- Diabetes affect more than **10%** of U.S. Population as of 2020
 - Type II diabetes account for over 90%
- 7.8 million hospital discharges of adults in U.S. had **diabetes** in their diagnosis in 2016
- 16 million **emergency visits** in 2016 had diabetes as a list of diagnosis
 - Over 10% of total 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 admission, <30 days, and >30 days

METHODS



Data Acquisition

Downloaded from
Kaggle based on
VCU research

Data Prep

Cleaned missing data

Recategorized features

EDA

Visualization

Statistical Test

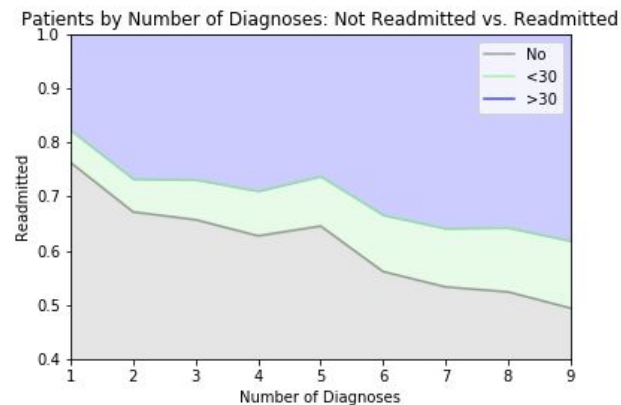
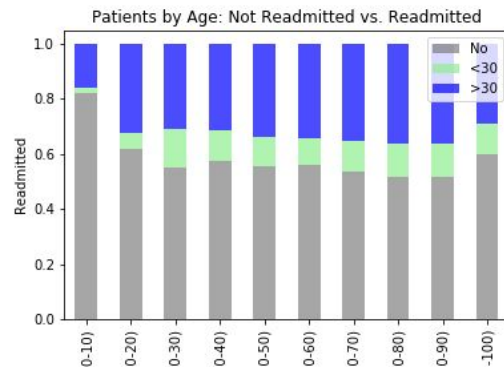
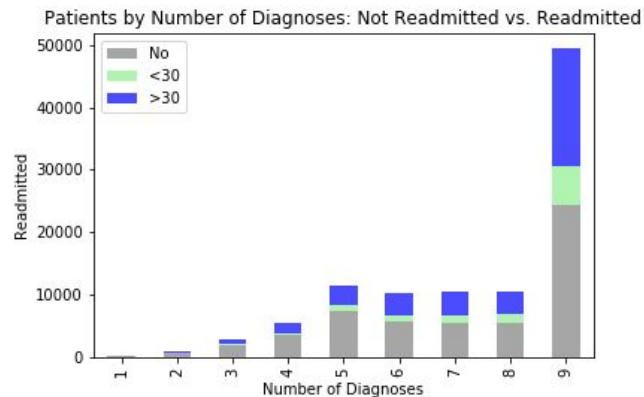
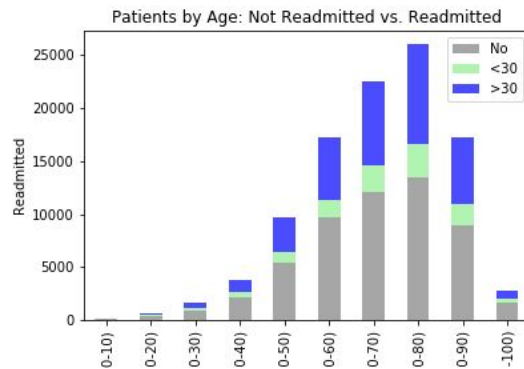
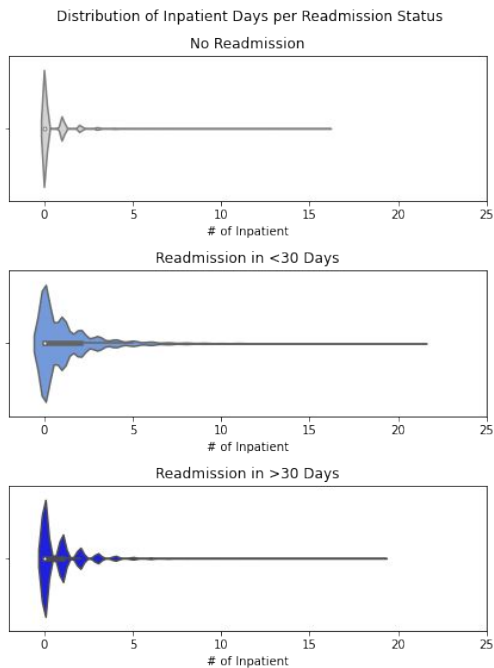
Models: Decision Tree, Random Forest

Metric: Recall, “micro” averaging

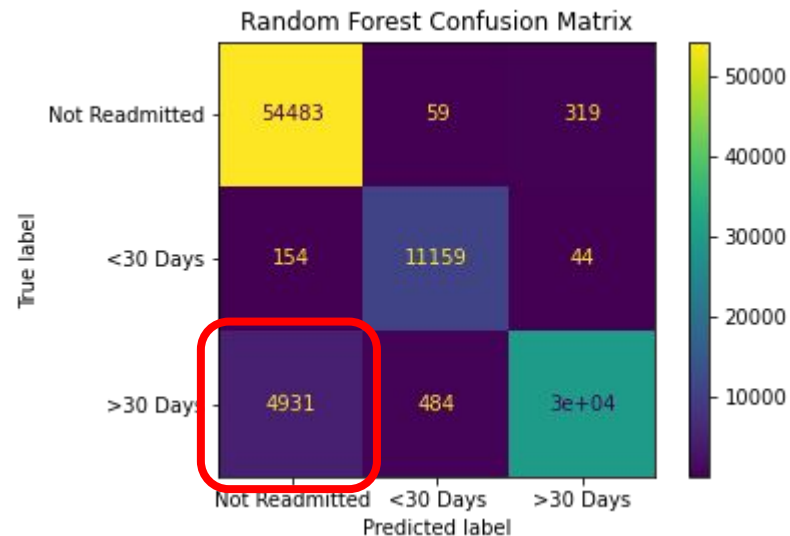
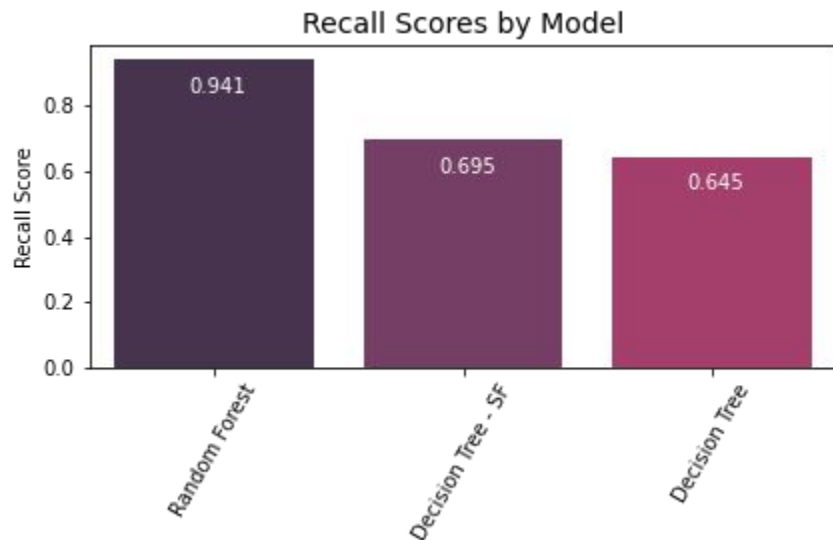
Hypertune: Grid Search Cross Validation

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](#)

