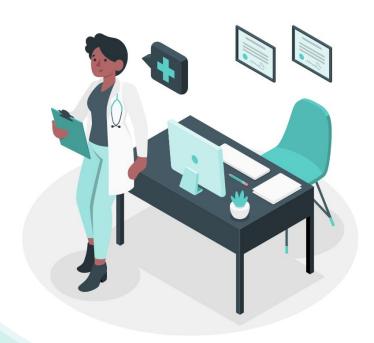
### Readmission Prediction of Diabetic Patients

Yung Han Jeong

Malcolm Katzenbach



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**02**Data

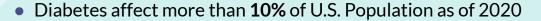
**O5**Conclusions

**O3**Methods

**06**Future Steps



#### 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**







Over 50 Features





## **METHODS**



**Data Acquisition** 

Data Prep

EDA

Downloaded from Kaggle based on VCU research Cleaned missing data

Visualization

Recategorized features

**Statistical Test** 

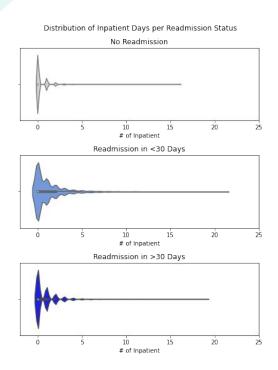
Models: Decision Tree, Random Forest

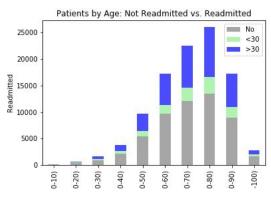
Metric: Recall, "micro" averaging

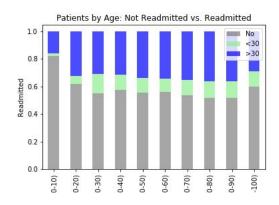
**Hypertune:** Grid Search Cross Validation

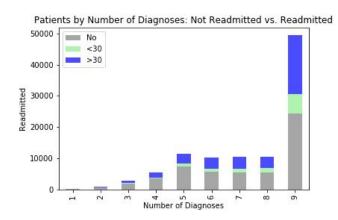
Score: Random Forest: 0.942

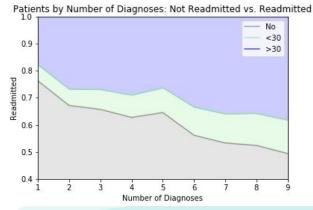
#### **MODELING - Features**



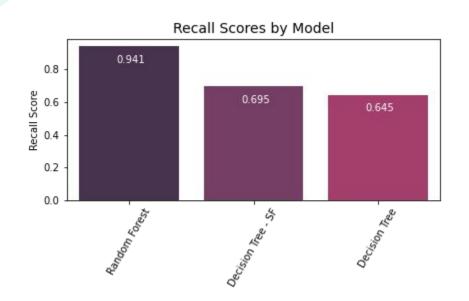


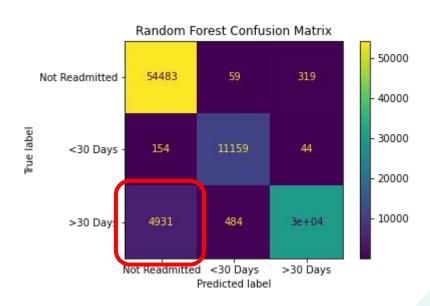






#### **MODELING - Performance**





#### **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

