

# Predicting Readmission of Diabetes Patients

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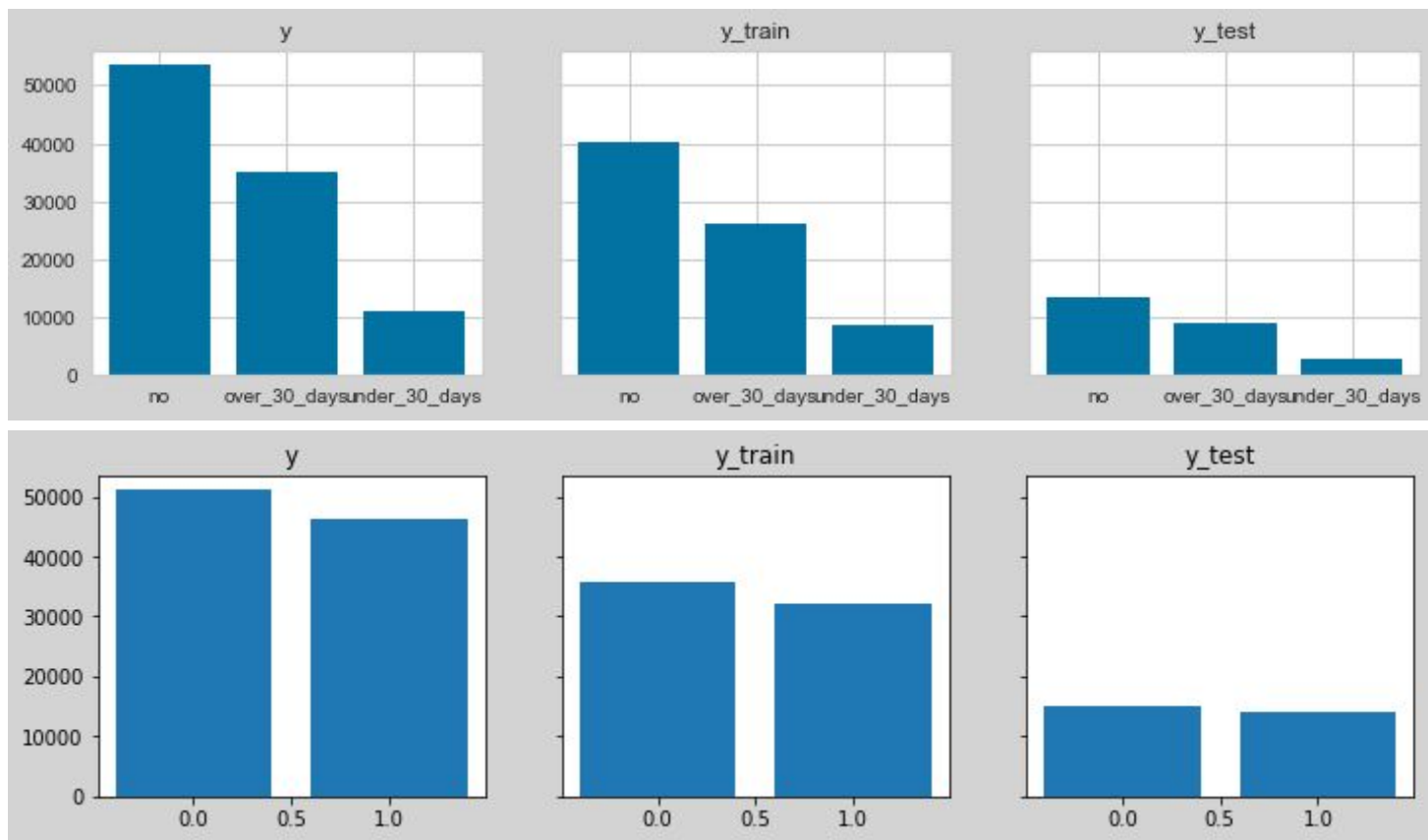


# Question & Dataset

In this project, we see if we are able to predict (at a rate better than purely random) the readmittance of diabetes patients given data both about the patient and the treatment the patient received.

The dataset we use is *Diabetes 130-US hospitals for years 1999-2008*, available in the Public Domain through [UCI's Machine Learning Repository](#) and [Kaggle](#). This dataset was first created for the open access journal article by Strack et al. "[Impact of HbA1c Measurement on Hospital Readmission Rates](#)", itself a subset of the **Health Facts Database** curated by the Cerner Corporation. The creation of this **100,000+ record dataset** was a query of diabetes diagnoses among **74,000,000 visitation records** meeting certain criteria (such as length of stay being 24 hours or greater, or laboratory examinations being performed on the patients).

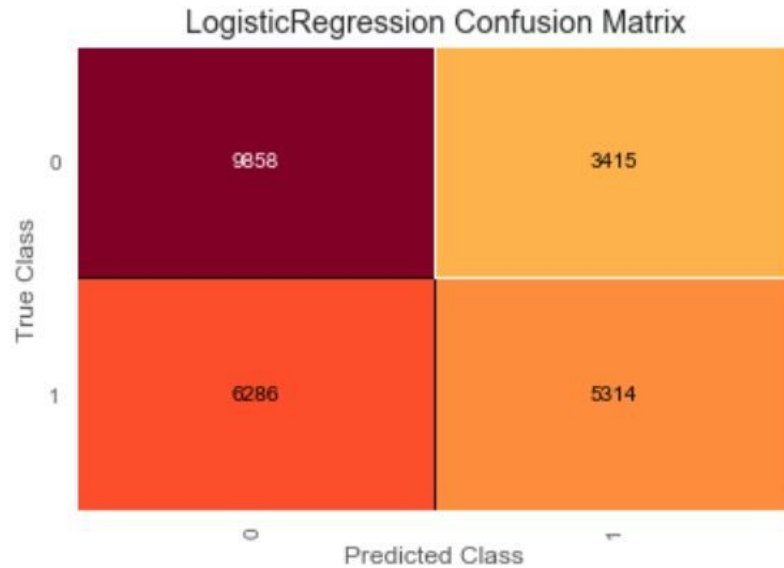
# Data Cleaning, Note #1: “Natural Balancing”



# Features

Demographics	Visitation Data	Diagnoses	Medication Data	Diabetes Specific Information	Testing
'female', 'Age', 'african_american', 'asian', 'hispanic', 'other',	'time_in_hospital', 'num_lab_procedures', 'num_procedures', 'num_medications', 'number_outpatient', 'number_emergency', 'number_inpatient', 'number_diagnoses', 'change',	<b>ICD #1-19</b>	'Num_meds', 'num_down', 'Num_up',  <b>Up, Down or Steady:</b> Metformin, Repaglinide, Glimepiride Glipizide Glyburide Pioglitazone Rosiglitazone	'Diabetesmed', 'glu_200_300', 'glu_over_300', 'Glu_norm', 'insulin_down', 'insulin_steady', 'insulin_up'	'a1_7_8', 'a1_norm', 'al_over_8',

# Data Cleaning, Note #2: “I Don’t See Dead People”



# Supervised Learning Techniques Used

- Dummy
- Decision Tree
- Random Forest
- Logistic Regression
- K-Nearest Neighbors
- XGBoost

# ...a peek inside the “black box”

