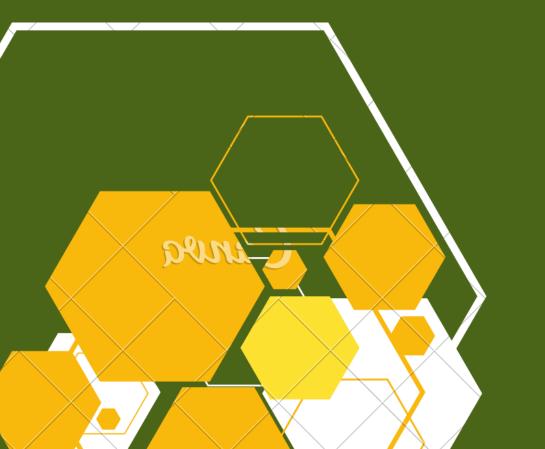


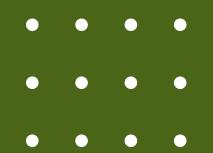




Fraud Detection Medicare Claims

Exploring ALTERYX

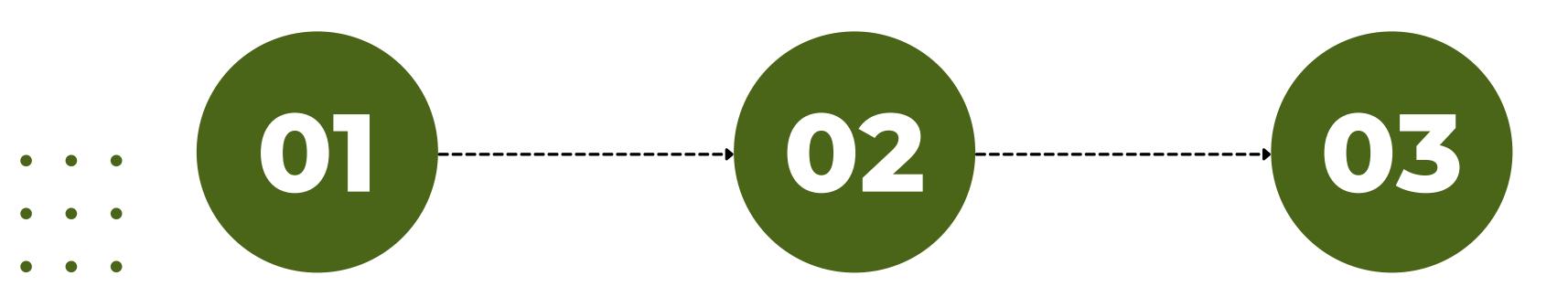




Detailed Overview

Fraud Detection using Medicare Claims dataset

- CMS dataset(M): Medicare Provider Utilization & Payment Data
- LEIE Dataset(L) List of fraudulent providers
- Goal With the two datasets identify the patterns in fraudulent claims and develop a classifier model.



Ingestion

Load all the data
Filter only the valid NPI data from L
Take Union of all the M datasets
Inner Join with NPI: Left join on L and Right join on M
Define Fraud Label with Y/N
Keep only non-null unique values & save as csv

Curation

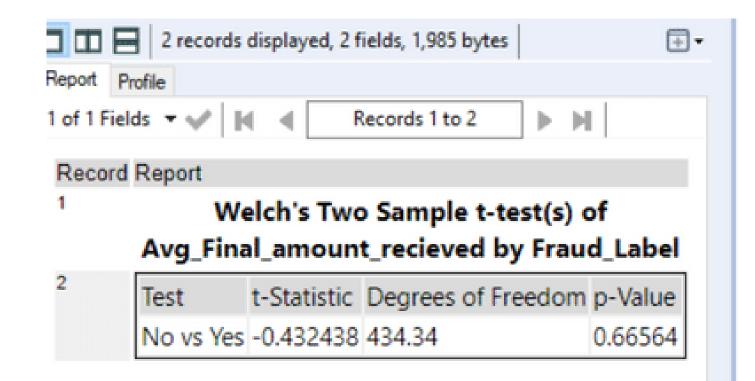
Use the ingested data
Only Unique NPIs & HCPCS code
Convert to respective datatypes
Define & add some custom metrics
Summarize the values with group by, count & average
Save the output data as csv file

ML Model

Performed t-test (Num var.)
Contingency Table (Cat var.)
Tried Logistic Regression &
Boosted Model

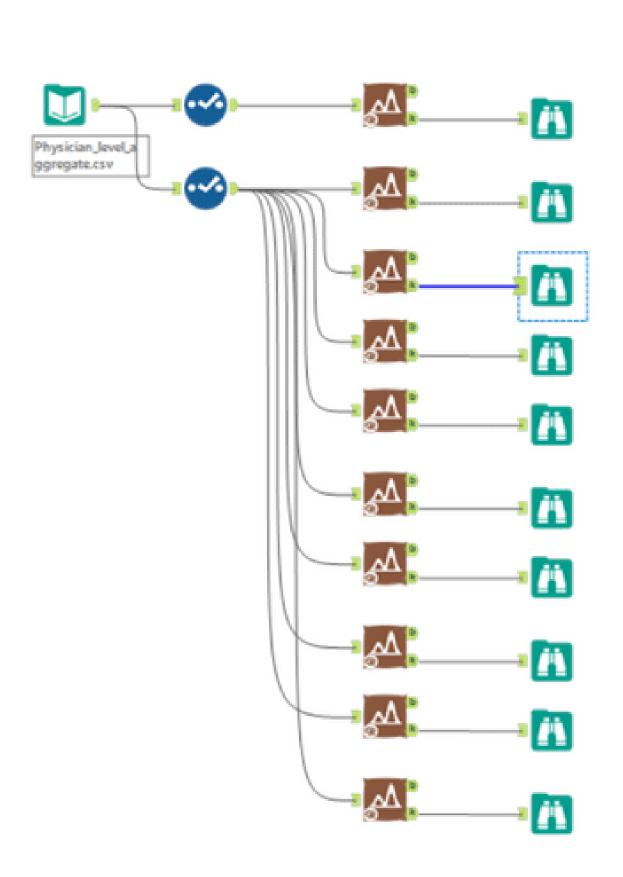
Statistics Data Report

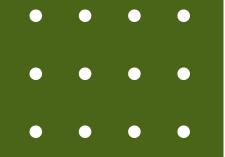




Numerical Variables: take only those where pvalues < 0.05



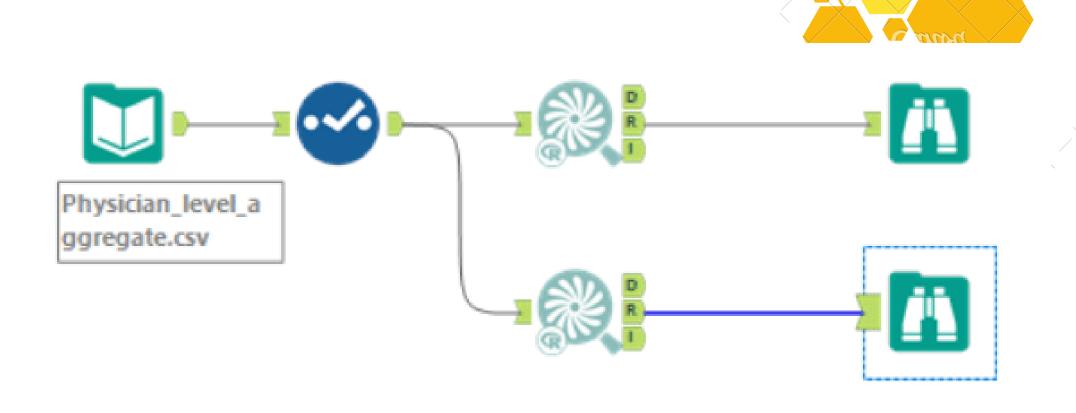




Select the Categorical Variables

Used contingency table Refer to p values





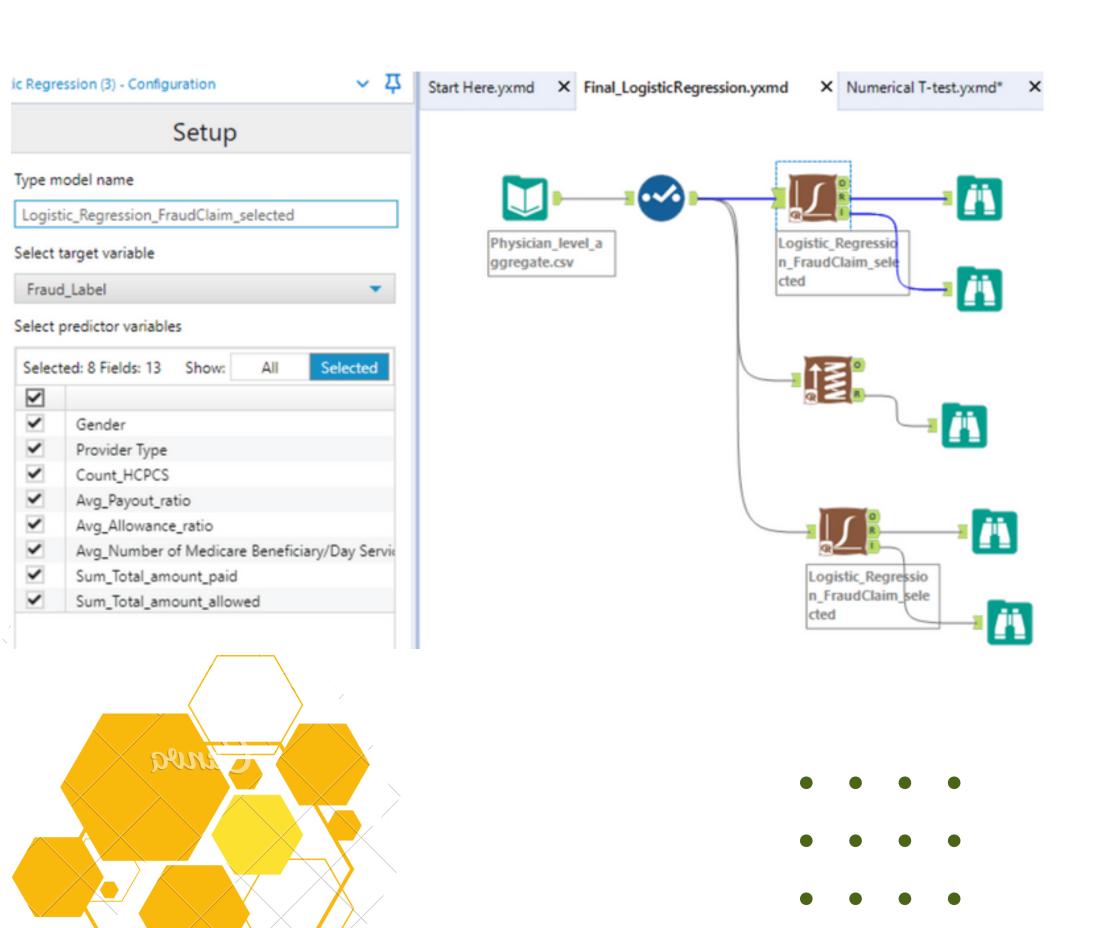
Provider Type & Target Value(Fraud _Label)

Chi-squared = 184.5167, df = 62, p-value < 0.0000

Gender & Target Value(Fraud _Label)

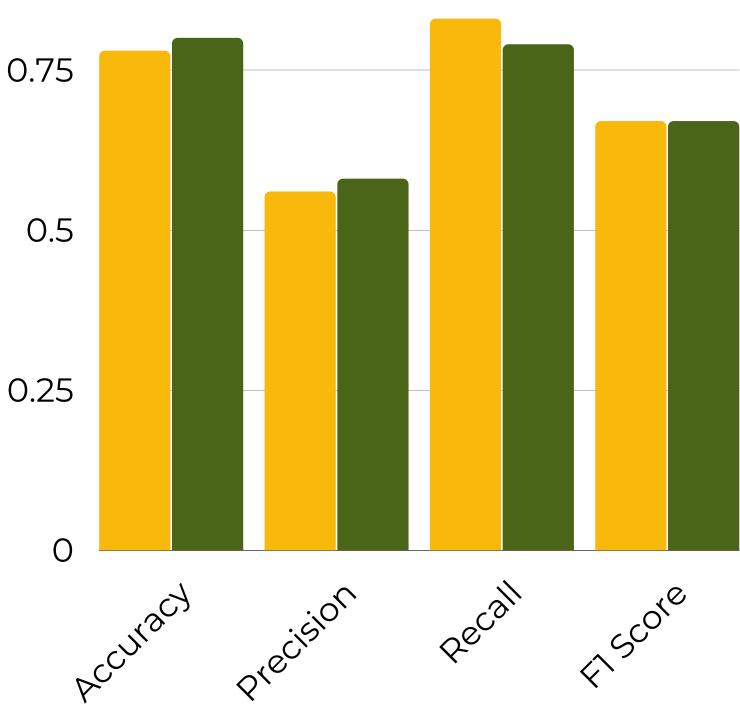
Chi-squared = 101.1603, df = 2, p-value < 0.0000

Compare the Classifier



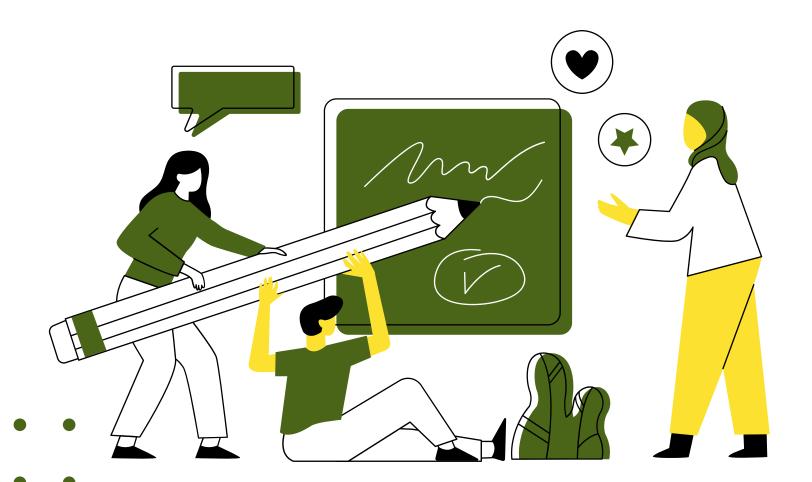
FP is more important - Precision





Feature Importance

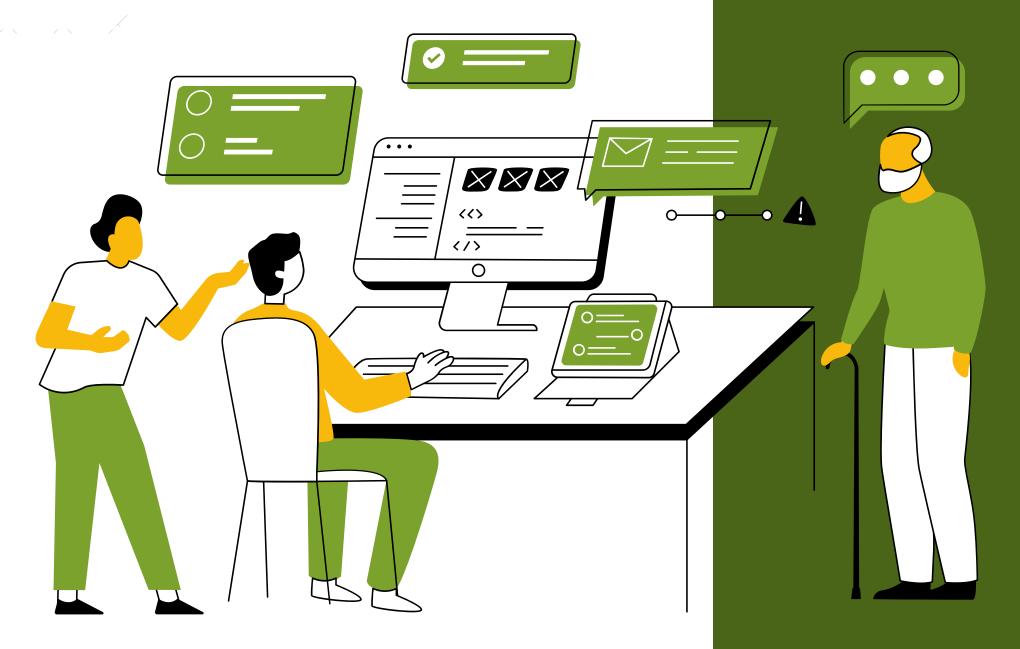
From Boosted Model (Same Table & Features as previous)



Provider.Type Gender Sum_Total_amount_allowed Avg_Number.of.Medicare.Beneficiary.Day.Services Avg_Allowance_ratio Sum_Total_amount_paid Count_HCPCS Relative Importance

Variable Importance Plot provides information about the relative importance





Further Steps

Lasso/Ridge regression
Have the Model with only Imp
Features

