



SEASONAL FLU VACCINE UPTAKE PREDICTION

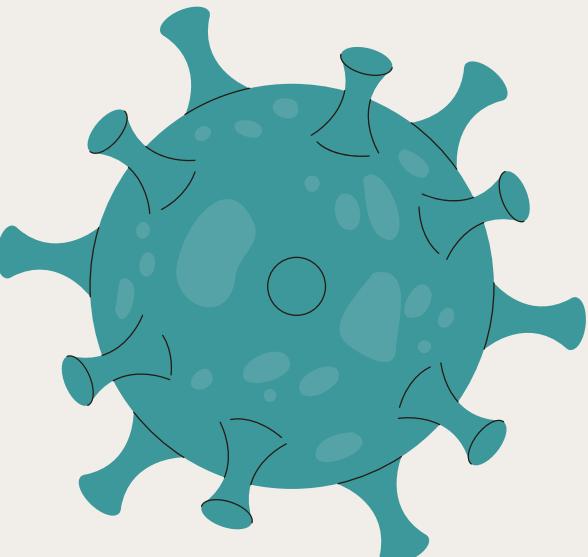
BY RONALD ONGORI NYAGAKA

INTRODUCTION



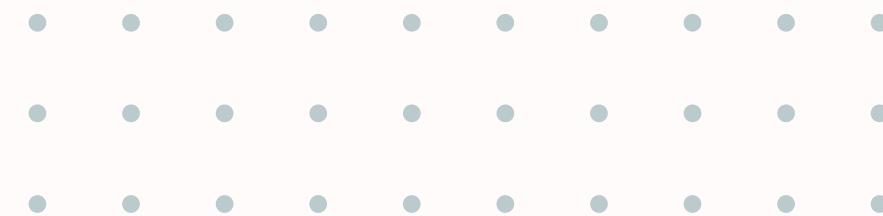
The occurrence of the flu season repeats annually, and each year people make a choice to either receive the flu shot or not.

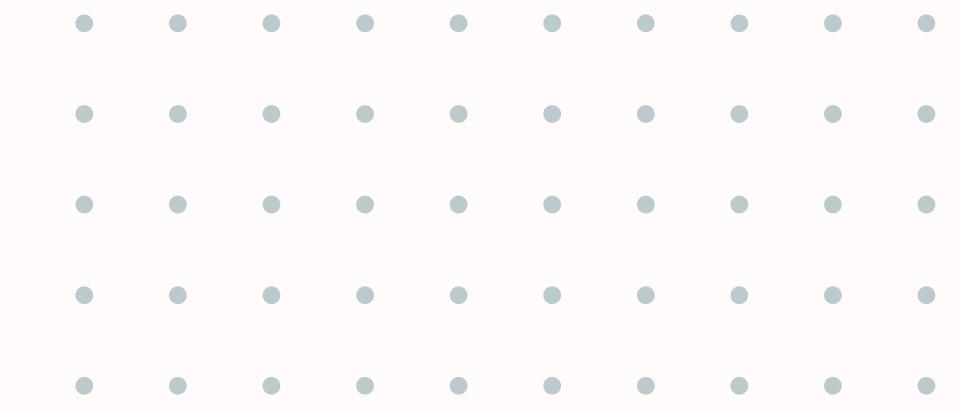
This attempts to develop a predictive model to forecast individuals' decision to receive the flu shot or not during the annual flu season.



PROBLEM STATEMENT

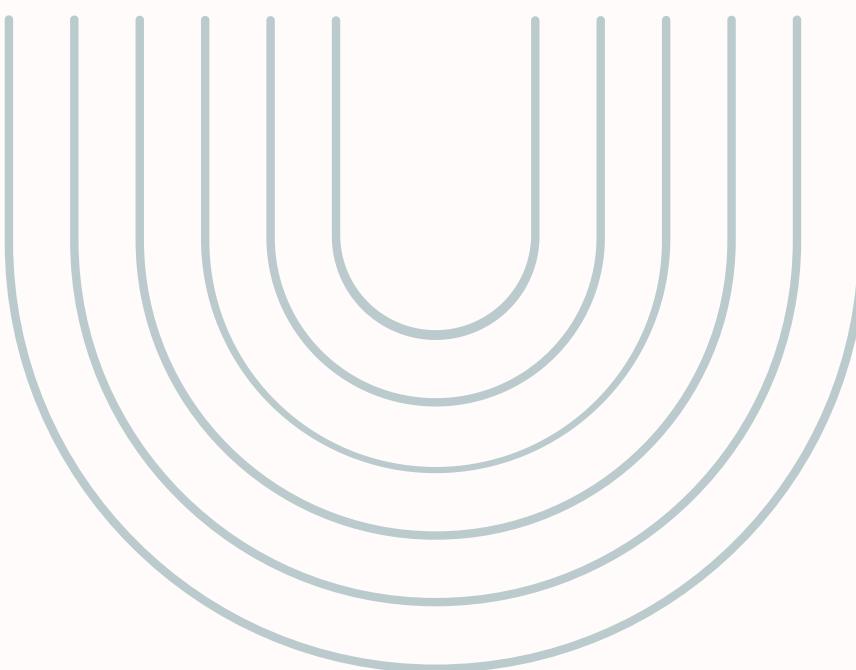
The aim of the project is to assist in public health planning and decision-making by providing insights into vaccination trends and helping allocate resources effectively to the United States government health Agencies





MAIN OBJECTIVE

Create a model that can predict seasonal flu vaccine uptake based on a person's background and patterns of behaviour



IMPORTANT FEATURES THAT SHAPE OUR MODEL



OPINION_SEAS_RISK

Respondent's opinion
about risk of getting
sick with seasonal flu
without vaccine

DOCTOR_RECC_SEASONAL

Seasonal flu vaccine
recommendation by
daughter

AGE_GROUP

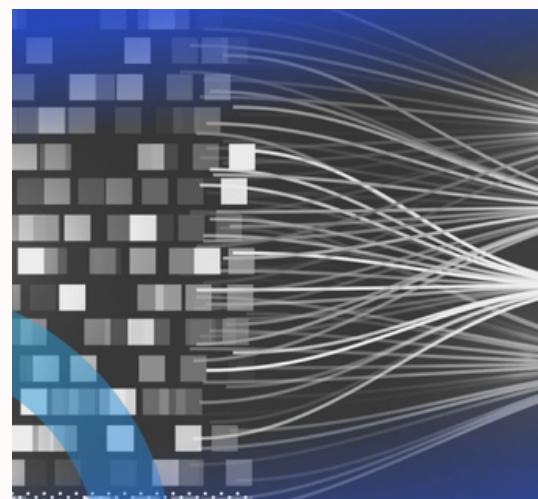
Age group of
respondents

EMPLOYMENT_INDUSTRY

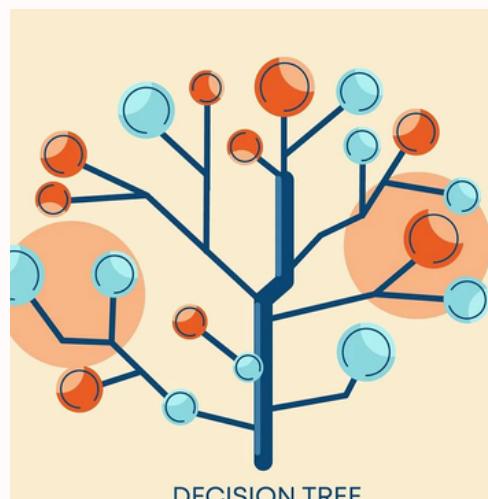
Type of industry
respondent is employed
in

MODELLING

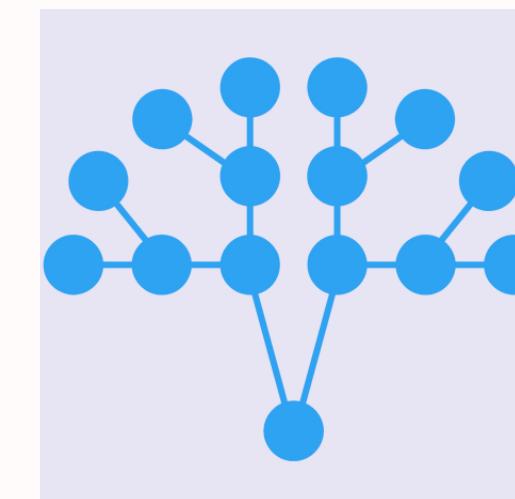
The baseline model was created using a Decision Tree Model. It had the lowest accuracy of 73.1 %



ADABoost



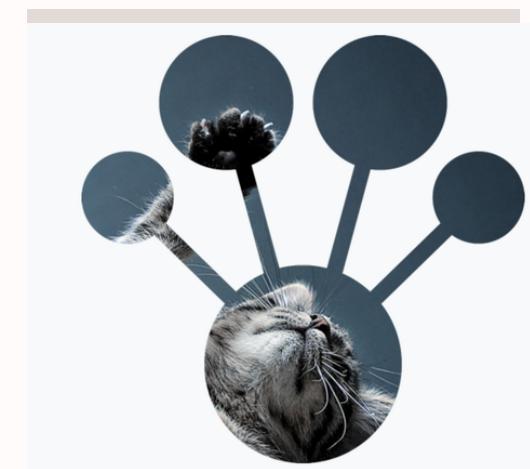
RandomForest



GradientBoost

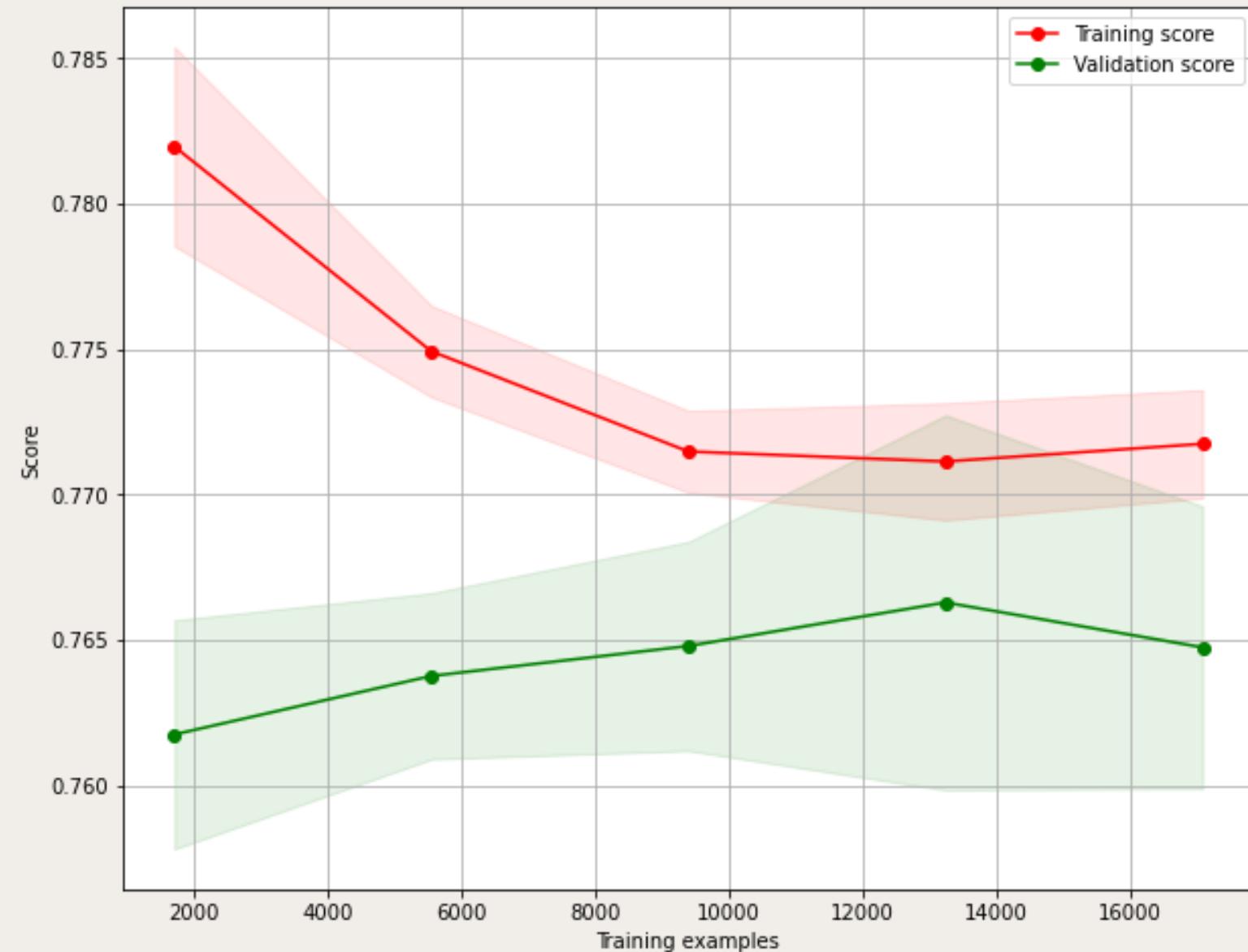


XGBoost

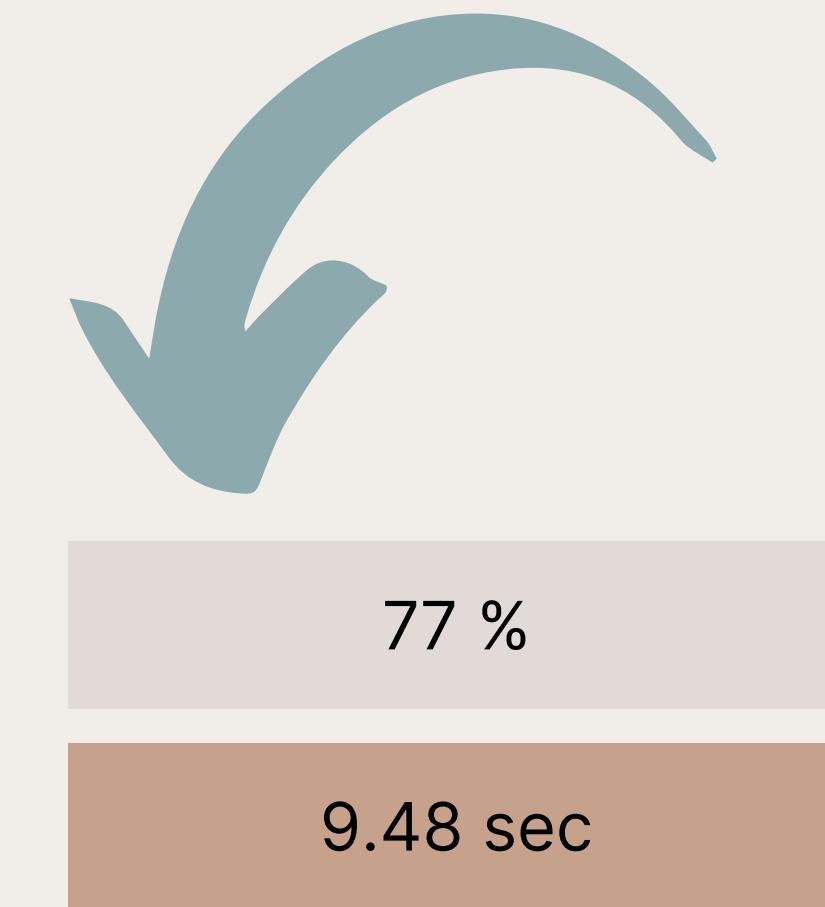


CATboost

RANDOM FOREST

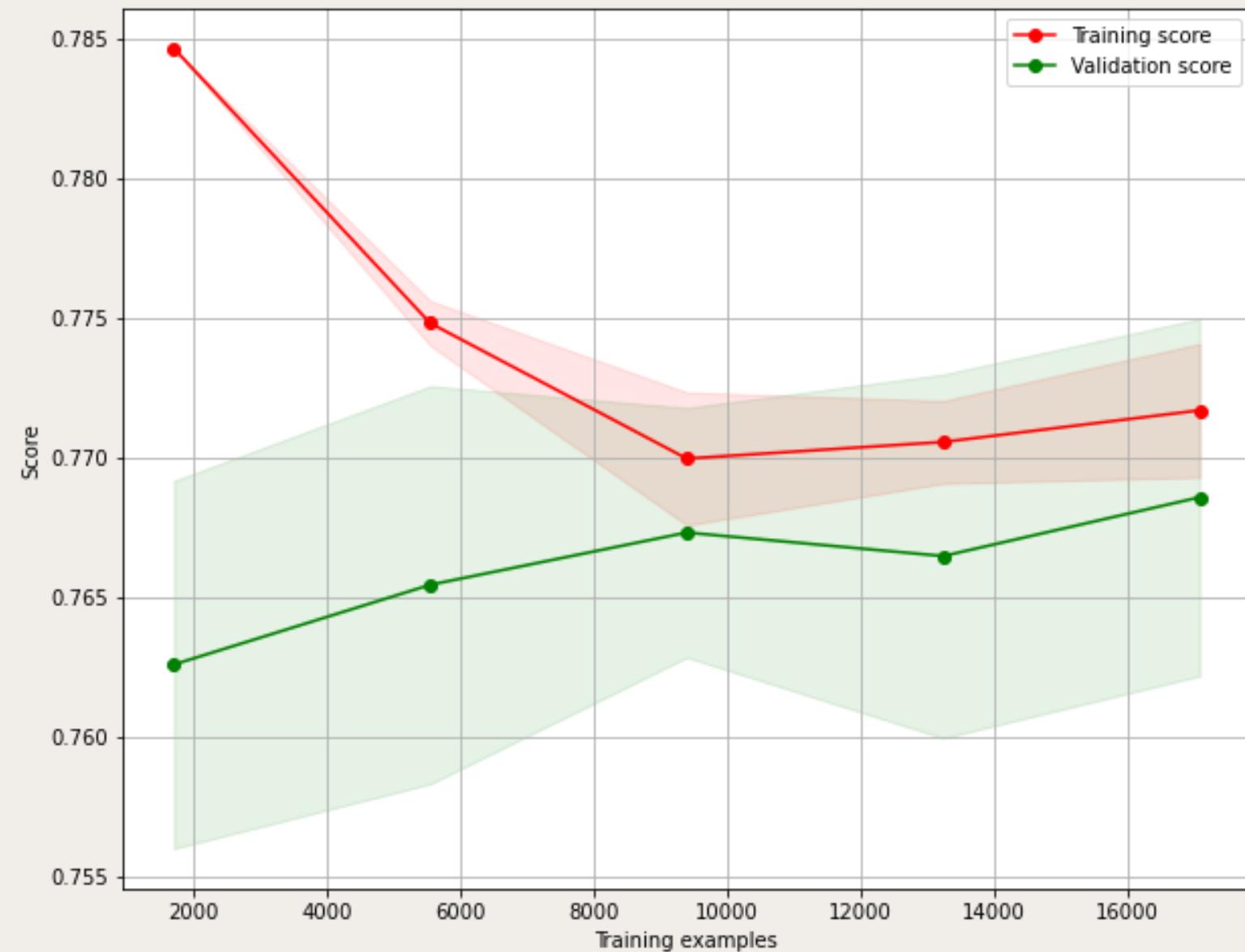


MODEL
ACCURACY
SPEED

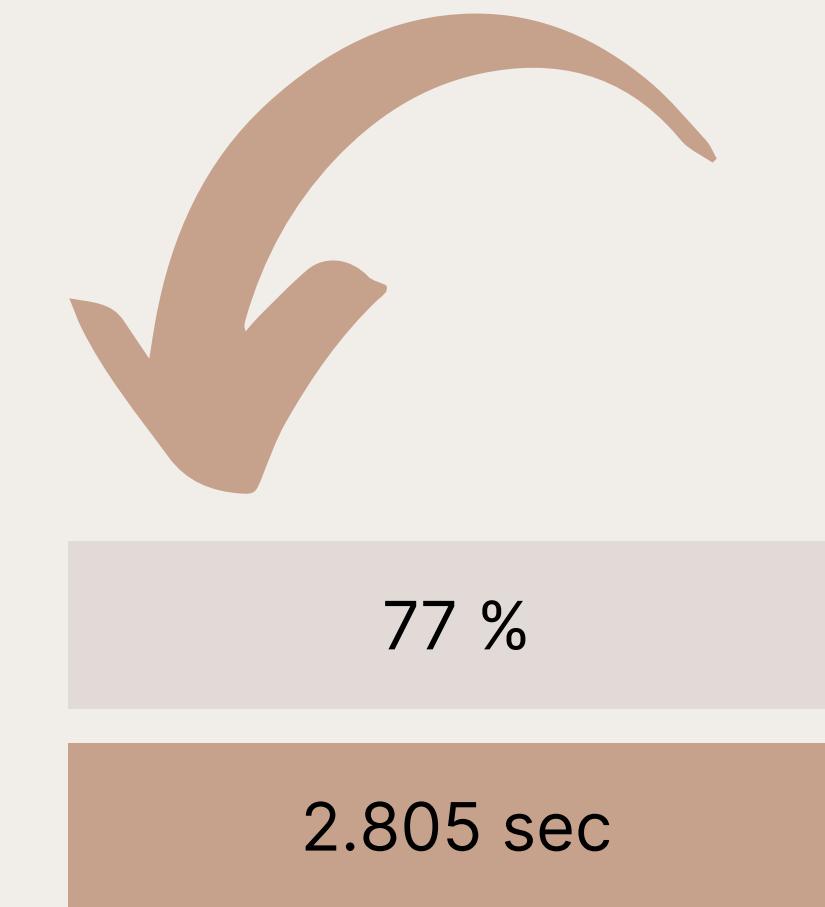


test score eventually plateaus and starts to decline, suggesting limitations in capturing complex patterns

GRADIENT BOOSTING

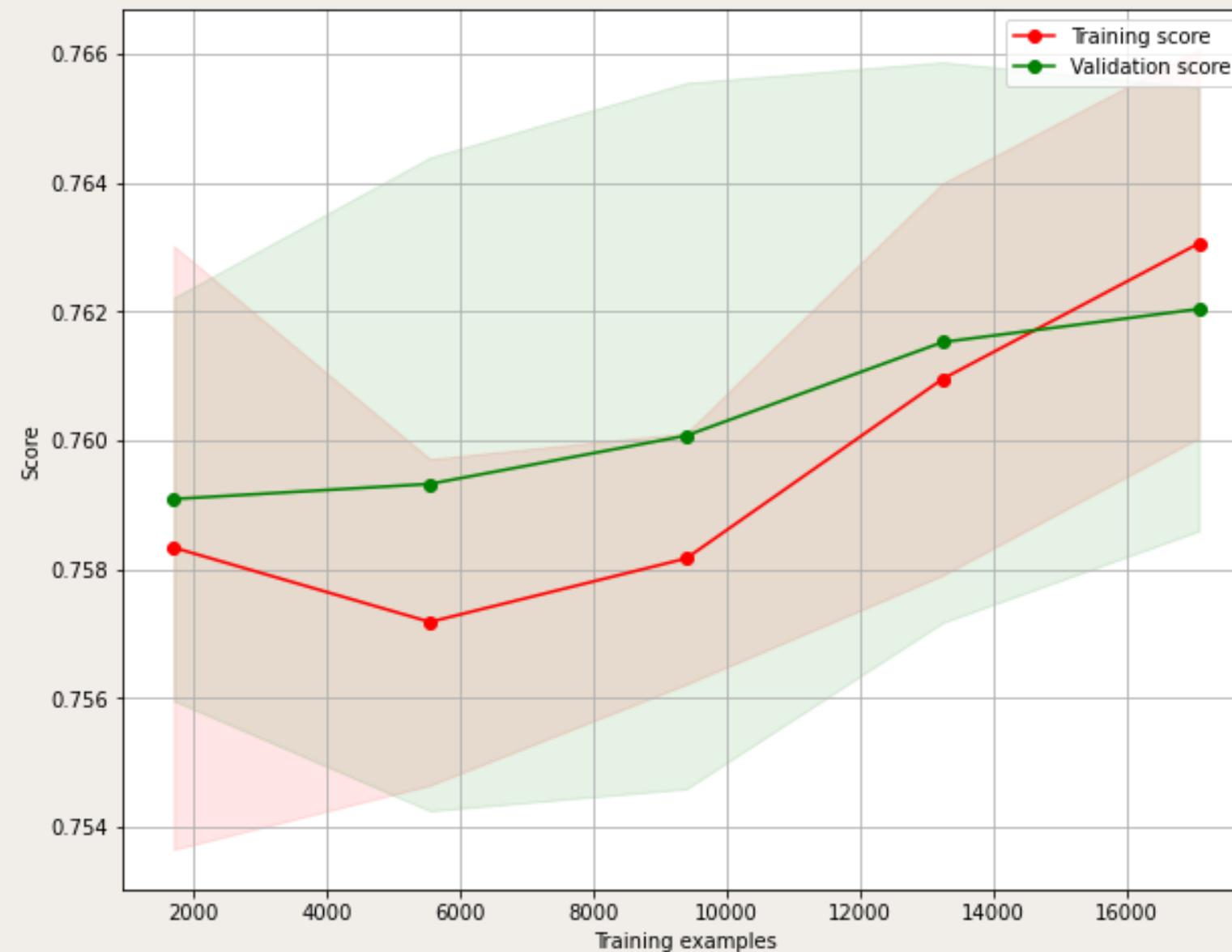


MODEL
ACCURACY
SPEED

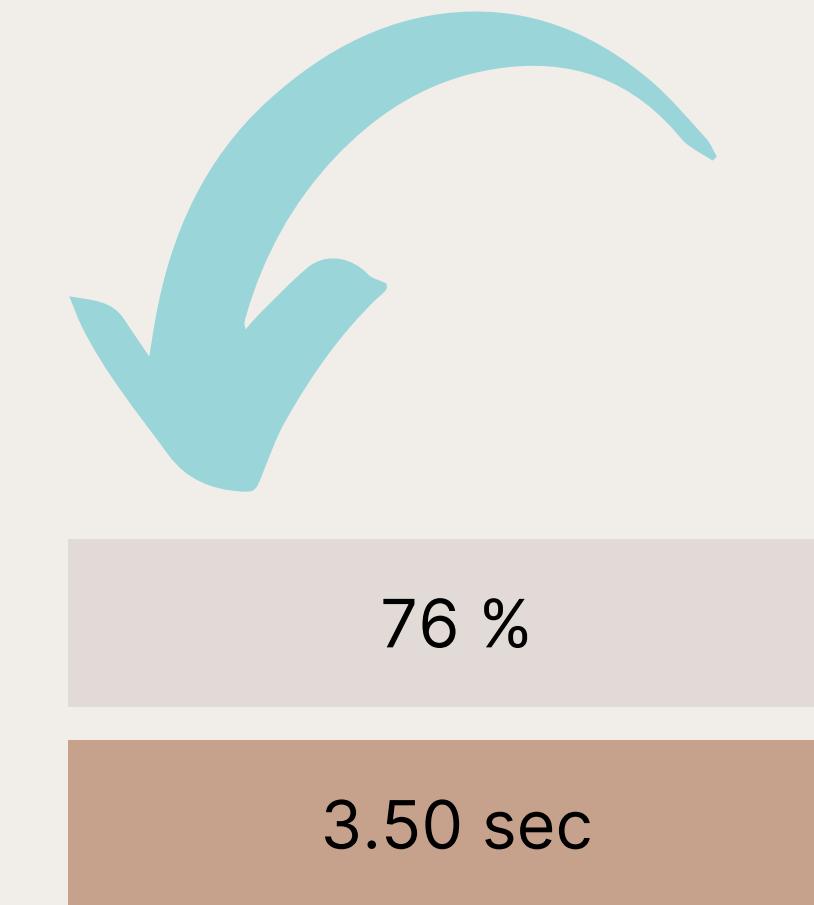


Can adjust and improve performance and generalizes well on unseen data. could improve if trained on more data

ADABOOSTING

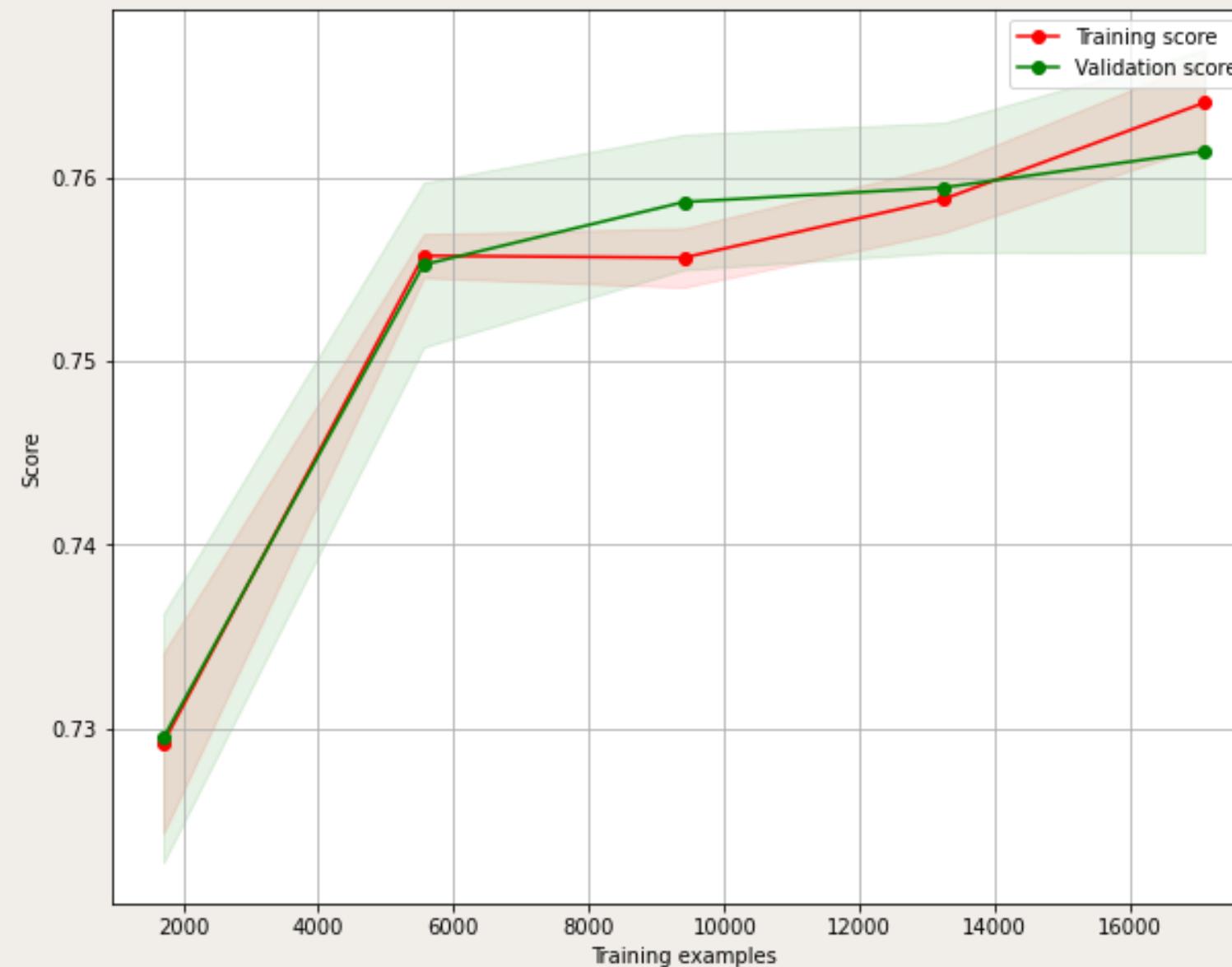


MODEL
ACCURACY
SPEED

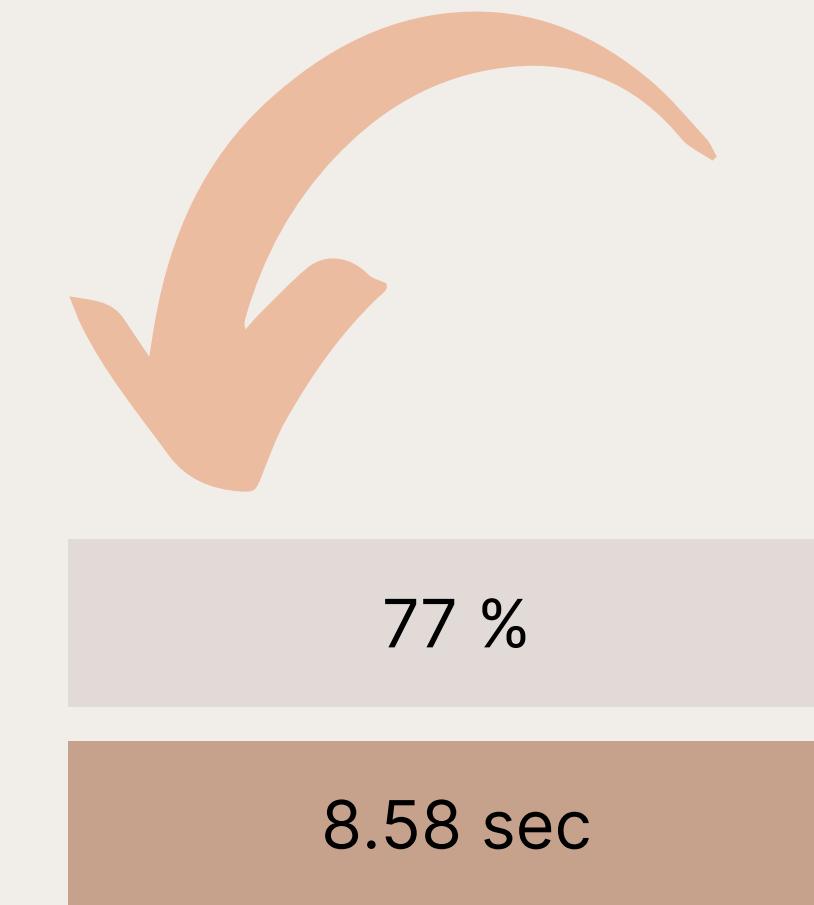


Model struggles to fit data initially but gradually improves. However its prediction ability starts to decline after some point

XGBOOST

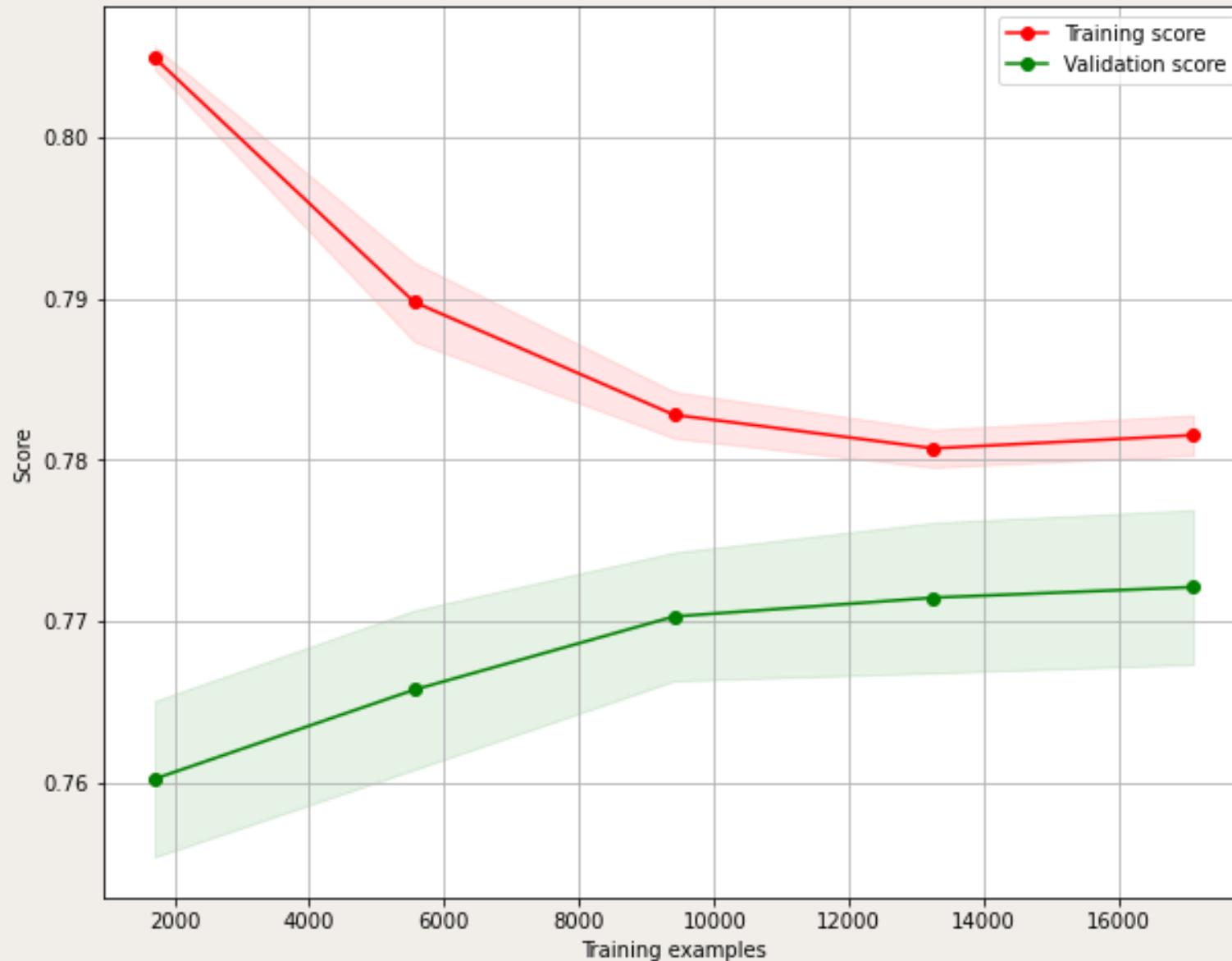


MODEL
ACCURACY
SPEED



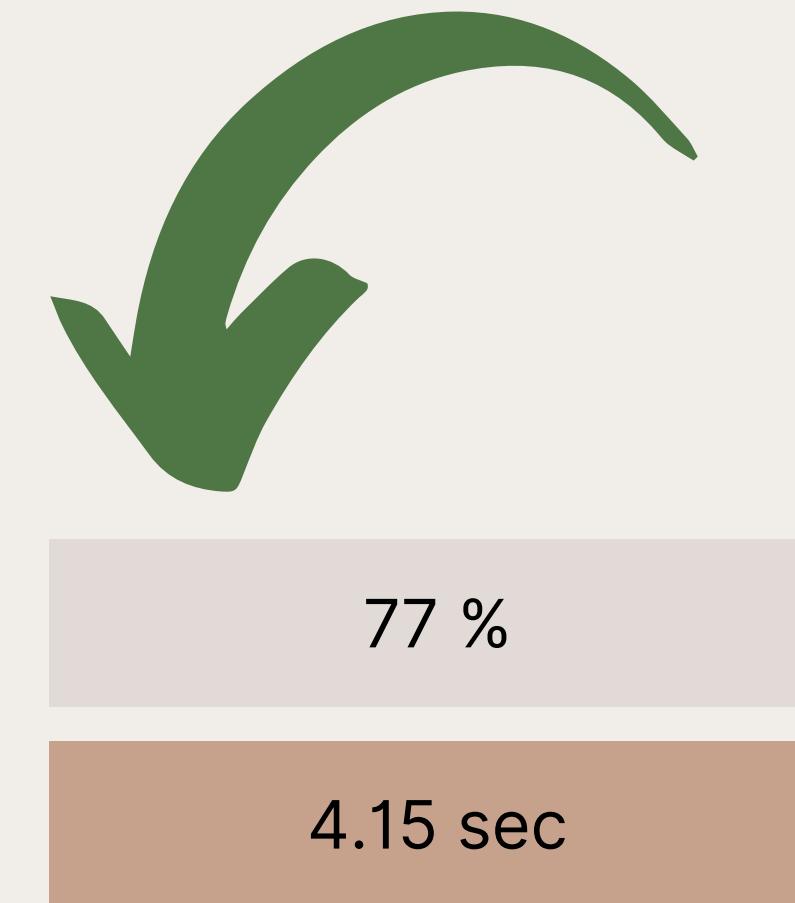
model benefits from additional training and
has the potential to further improve its
accuracy. However it starts to overfit

CATBOOST



MODEL
ACCURACY
SPEED

model is overfitting and shows signs of a
plateauing performance as more data is
trained



THE BEST MODEL: GRADIENT BOOSTING

the validation score rises steadily together with the training score as more data is trained. It is improving its predictive accuracy and it adjusts to improve performance on challenging instances.

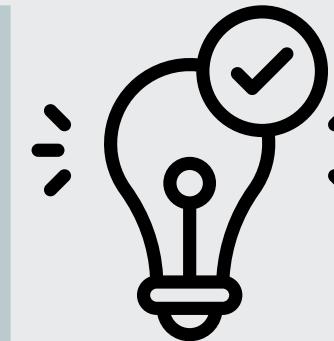
Gradient boost has better prospects on predictive performance as more data is introduced



CONCLUSION



The data set was biased towards certain communities and classes.



This makes the models generated biased towards this group of people



Better insights would be generated if the data was more balanced.



Recommendation



- Communication of limitations, biases, and specific applicability of the models, Should be adhered to.
- Resampling techniques such as oversampling, undersampling can be employed to mitigate biase.
- Careful examination of the significance of identified predictors will help understand underlying factors and ensure fairness in decision-making
- Include a wider range of demographic groups, thereby achieving a more diverse and balanced dataset.