# Social Pressure and Voting

# Overview

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- 03. Advanced EDA
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# The Problem

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Low Voter Turnout

2

Poor Political Representation

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How might we use machine learning to understand the role of social pressure in determining voting behaviour

such that we can design interventions to increase voter turnout?

dentis

subgroups

#### Dataset Overview

Target column is 'voted'.

G 2 0 0 0 P 2 0 0 0 D E M

Past Voting Behaviour

S E X Y O B % < 5

Age/Gender

MEDIAN \$ UNEMPLOY

Income & Employment

> H S > B A C H < 9 - 1 2

Education

ZIP HH\_ID

Location & Household Info.

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### Feature Engineering

1

#### **One-Hot Encoding**

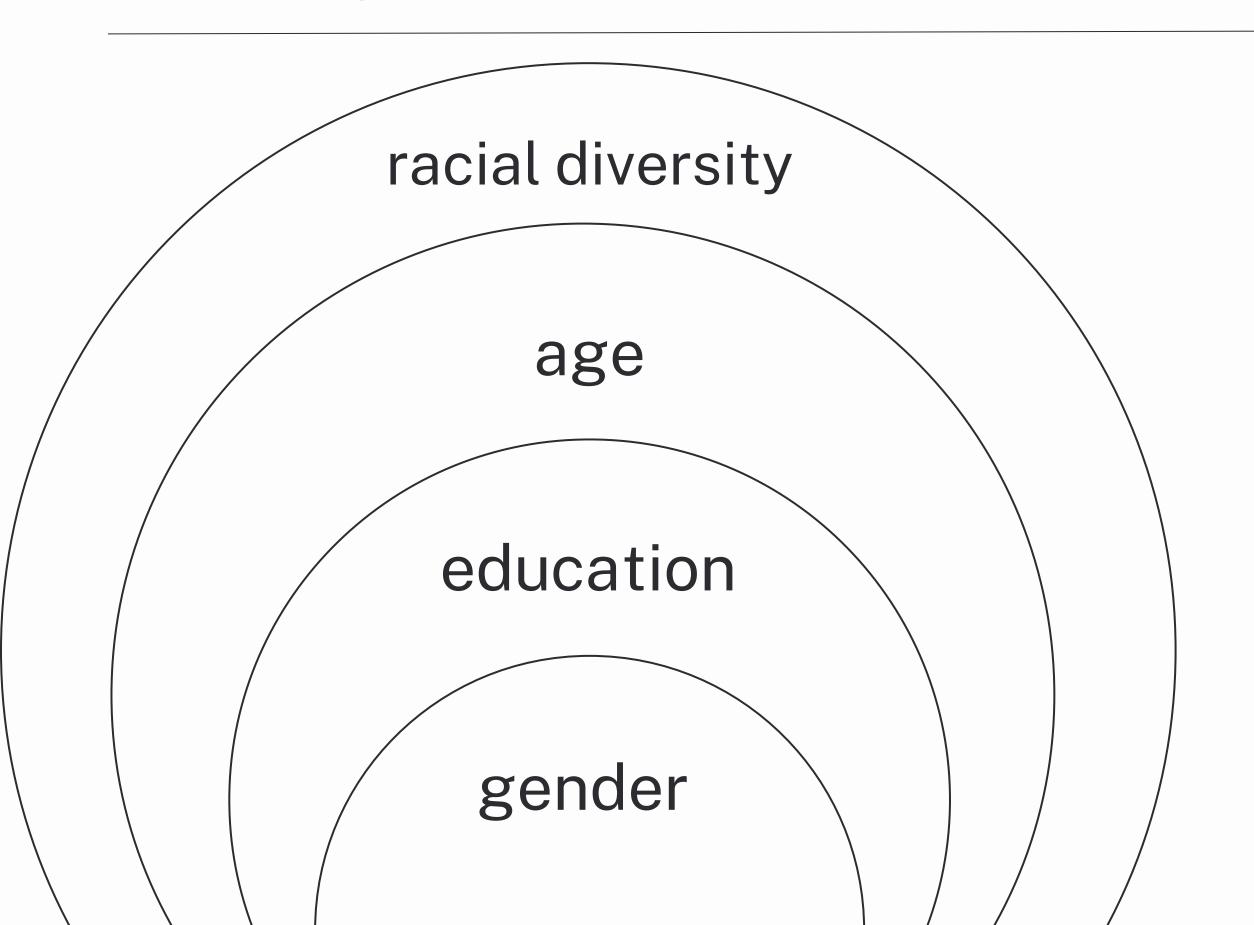
Turned treatment columns into dummy columns.

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#### **Dropping Columns**

Removed columns that were redundant, had low variance or were highly correlated with other variables.

#### ADVANCED EDA



#### **CLUSTERING**

Using MiniBatchKMeans

#### **SCORING**

LUsed Inertia and Silhouette Score.

#### **MODELLING**

Yet to be integrated into modelling but promising.

#### MODELLING APPROACHES

## Logistic Regression

- 7 models
- Generally, accuracy ~69%
- Good precision and recall for the majority class (not voted)
- Poor precision and recall for minority class
- Used SMOTE

## Decision Trees

 Hyperparameter tuning using RandomizedSearchCV and GridSearchCV

#### Random Forest

- 2 models
- Better recall for 'first pass' model with no drop in accuracy
- Much better recall for class-balanced model

## **Gradient Boosting**

• Slight improvement in accuracy and precision

## Neural Network

- 3 layers, funnel-shaped
- Used dropout layers for generalisability

## Evaluation Metrics

Accuracy

F1 Score

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#### NEXT STEPS

Clustering

Integrate into modelling

**Causal Forests** 

Estimate CATE

**Class-Balancing** 

Try with more models

**Shapley Values** 

To interpret models better