



DATA³

KISS n DATA SCIENCE

HELEN TANNER

Hello...



UNIVERSITY OF
Southampton

We're a data lab...



DATA³

We help businesses to make more money using data they already have



KISS

keep.it.simple.stupid.

EXAMPLE 1 – INSURANCE BUSINESS

CUSTOMER VALUE METRIC

LIFETIME CUSTOMER VALUE

$$\text{EVEB} = \frac{P_1 \times (1 - C_1)}{(1 + r)} + \frac{R_1 \times P_2 \times (1 - C_2)}{(1 + r)^2} + \frac{R_2 \times P_3 \times (1 - C_3)}{(1 + r)^3}$$

where EVEB = Expected Value of Existing Business

P_i = profit at time i = Premium – E(Loss) – Expense

C_i = probability of cancellation during period i

R_i = Probability of renewal at the end of the period i

r = discount rate

The value of a customer in an insurance institution may be computed using the formula below

$$CLV_i = -AC_i + \sum_t^T \left(r_{ti} \times \frac{(AR_{ti} + UR_{ti} + CR_{ti} + RV_{ti}) - (SC_{ti} + MC_{ti})}{(1+d)^t} - r_{ti}^{t-1} \times (1-r_{ti}) \times \frac{TC_i}{(1+d)^t} + r_{ti}^t \times \left\{ \frac{InfoV_{ti} + CoopV_{ti} + InnoV_{ti}}{(1+d)^t} \right\} \right)$$

where:

CLV – value of insurer's customer i [net current value of future earnings generated by that customer],

AC – insurer's cost to acquire customer i , r – retention rate of customer i in period t , AR – basic revenue

generated by customer i in period t , UR – up-selling revenue generated by customer i in period t , CR –

insurer's cross-selling revenue generated by customer i in period t , RV – referral value produced by

customer i in period t , MC – insurer's marketing costs to retain customer i in period t , SC – costs of sales

of products and of service provided to customer i in period t , TC – costs of terminating an insurer's

cooperation with customer i in period t , InfoV – value of information an insurer receives from customer i

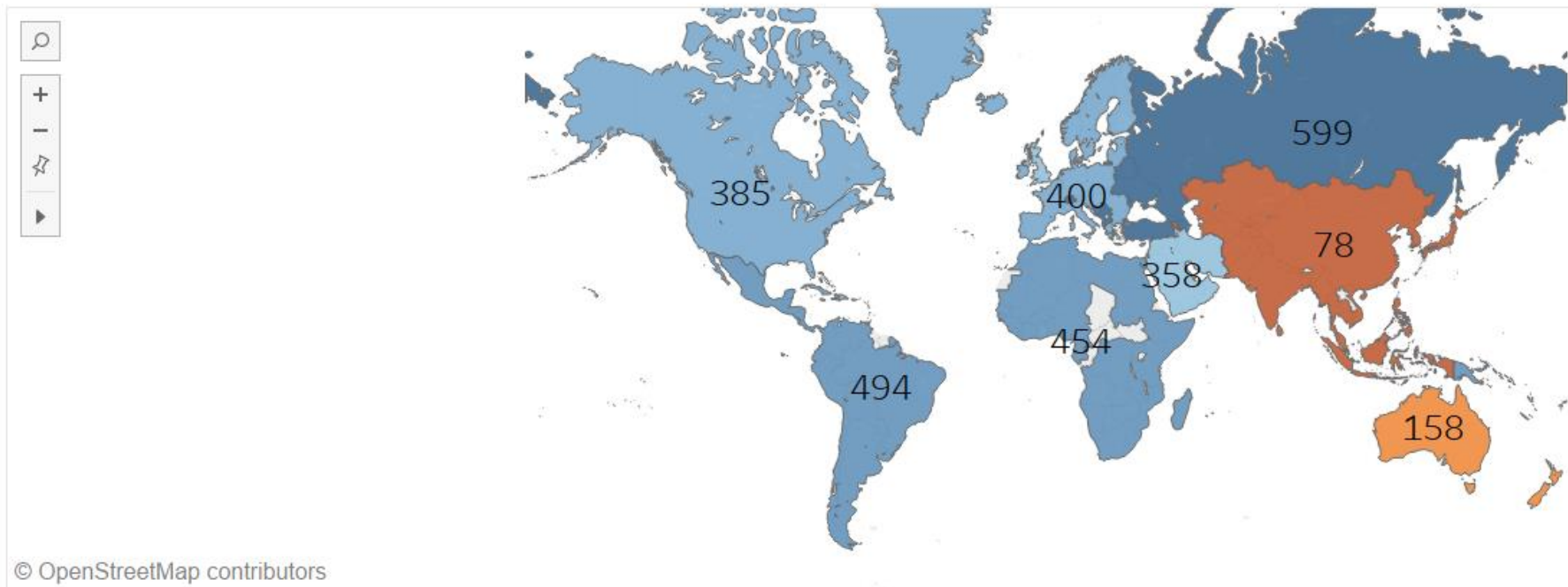
in period t , CoopV – value of an insurer's cooperation with customer i in period t , InnoV – value of innova-

tions proposed by customer i in period t , d – discount rate, T – number of periods considered (years).

$$\frac{\text{TOTAL PREMIUMS PAID} - \text{TOTAL COSTS \& CLAIMS}}{\text{NUMBER OF YEARS AS A CUSTOMER}} = \text{CUSTOMER VALUE PER CUSTOMER}$$

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Metric..



EXAMPLE 2 – RIDE HAILING BUSINESS

CUSTOMER VALUE METRIC

LIFETIME CUSTOMER VALUE

Annual Margin per Customer (m): the profit we will make on a sale after accounting for variable expenses (as opposed to just revenue)

Customer Retention Rate (r): the percentage of customers who continue to purchase in a subsequent year

Rate of Discount (i): the interest rate used to discount the value of future cash flows

$$LTV = m \left(\frac{r}{1 + i - r} \right)$$

$$\begin{array}{l} \text{Average} \\ \text{Revenue} \\ \text{per} \\ \text{Conversion} \end{array} \times \begin{array}{l} \text{Repeat} \\ \text{Rate} \\ \text{per} \\ \text{Consumer} \end{array} = \begin{array}{l} \text{LIFETIME} \\ \text{VALUE} \\ \text{OF} \\ \text{CONSUMER} \end{array}$$

TOTAL FARES x NUMBER OF FARES =

CUSTOMER VALUE PER PASSENGER

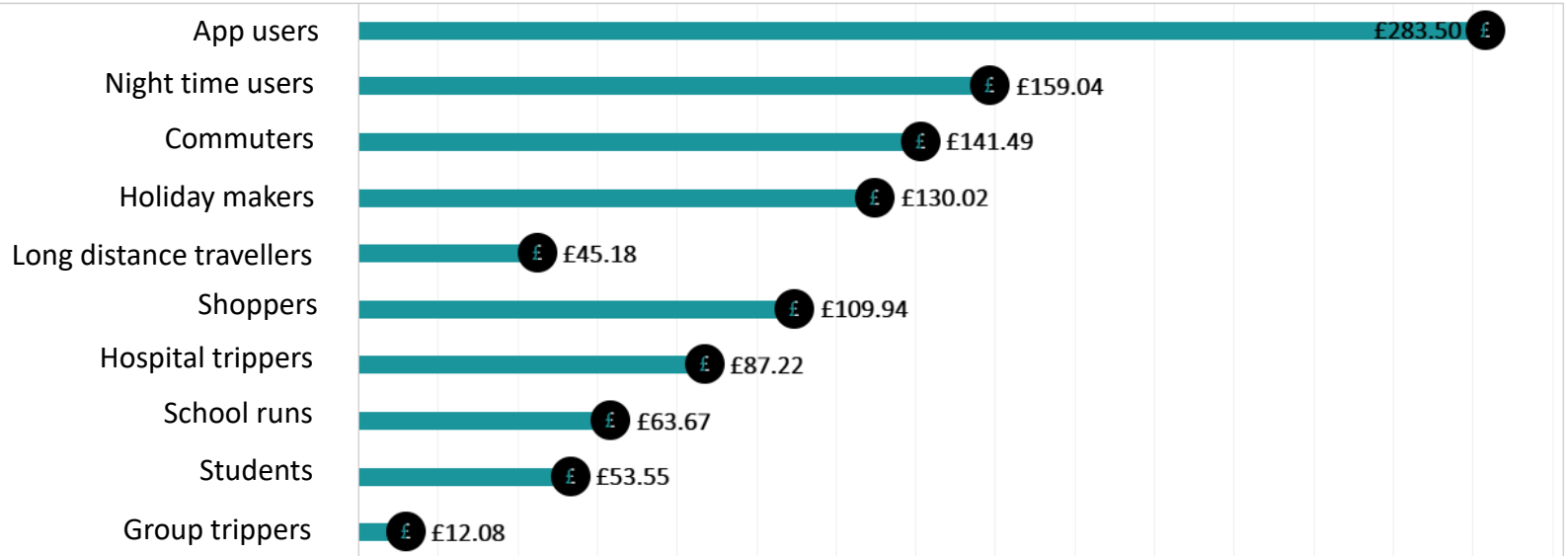
CUSTOMER VALUE PER DRIVER

TOTAL FARES x NUMBER OF FARES =

CUSTOMER VALUE PER PASSENGER

CUSTOMER VALUE PER DRIVER

Average customer value by frequency of trips, payment type and source



EXAMPLE 3 – EMPLOYEE BENEFITS BUSINESS

PREDICTIVE MODELS

PREDICTING EMPLOYEE BENEFIT PARTICIPATION

Data
Data
Data



Black box:

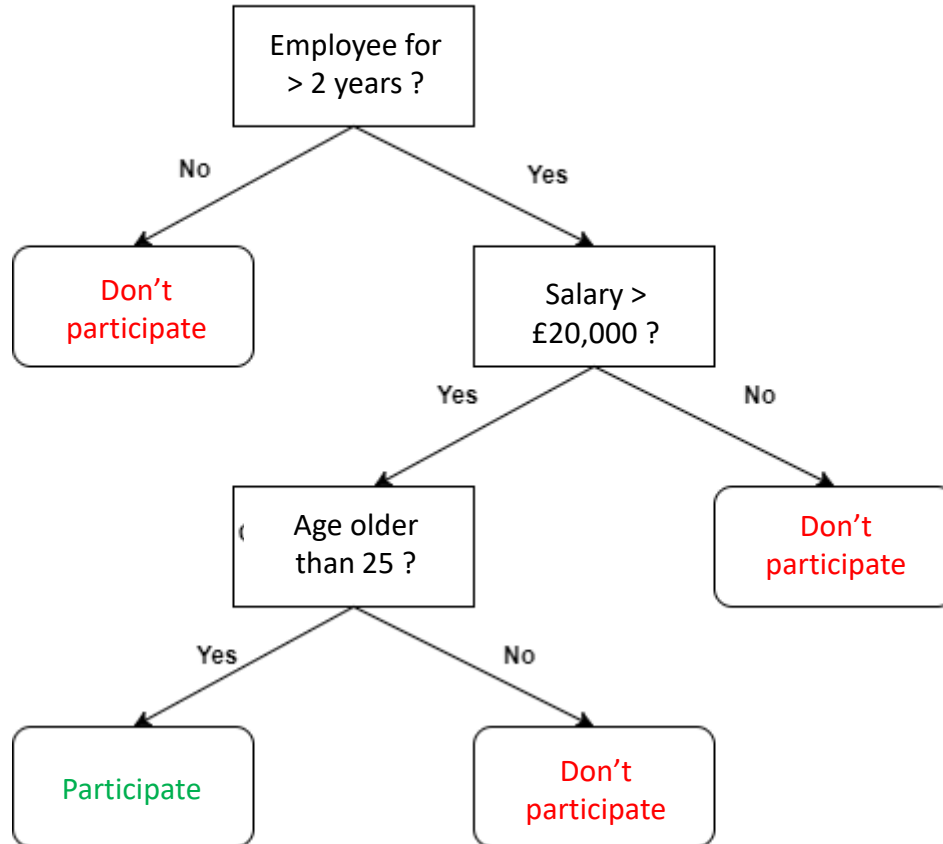
Statistical
procedures or
data science
algorithms



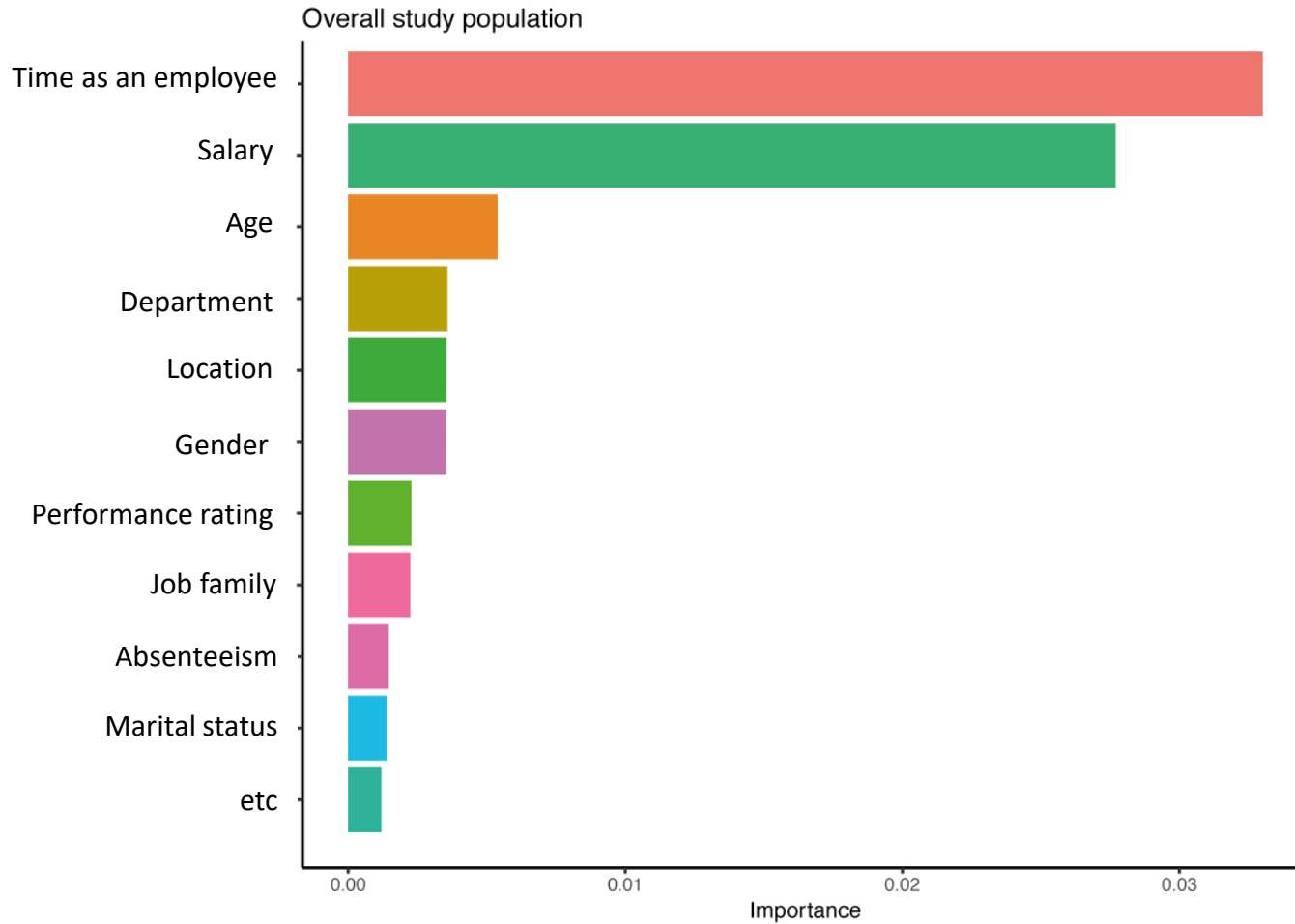
Results

PREDICTING EMPLOYEE BENEFIT PARTICIPATION

Decision Tree for Classification and Prediction



PREDICTING EMPLOYEE BENEFIT PARTICIPATION



EXAMPLE 4 – SHAREHOLDER VOTING

PREDICTIVE MODELS

PREDICTING SHAREHOLDER VOTING

Data
Data
Data



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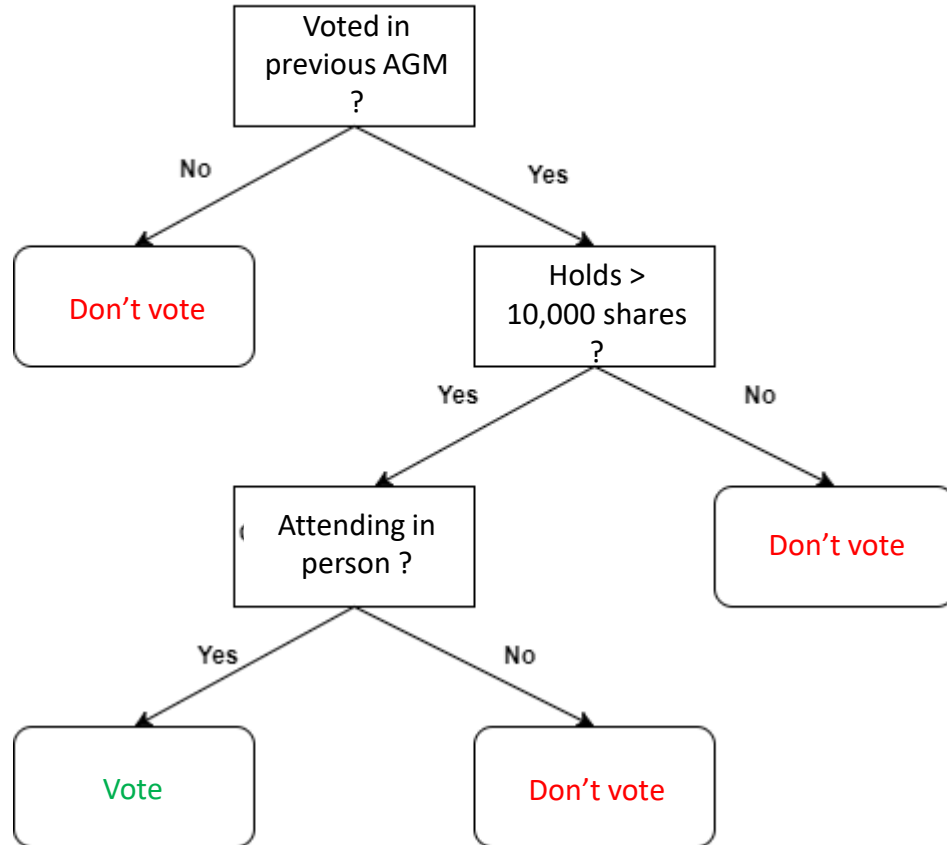
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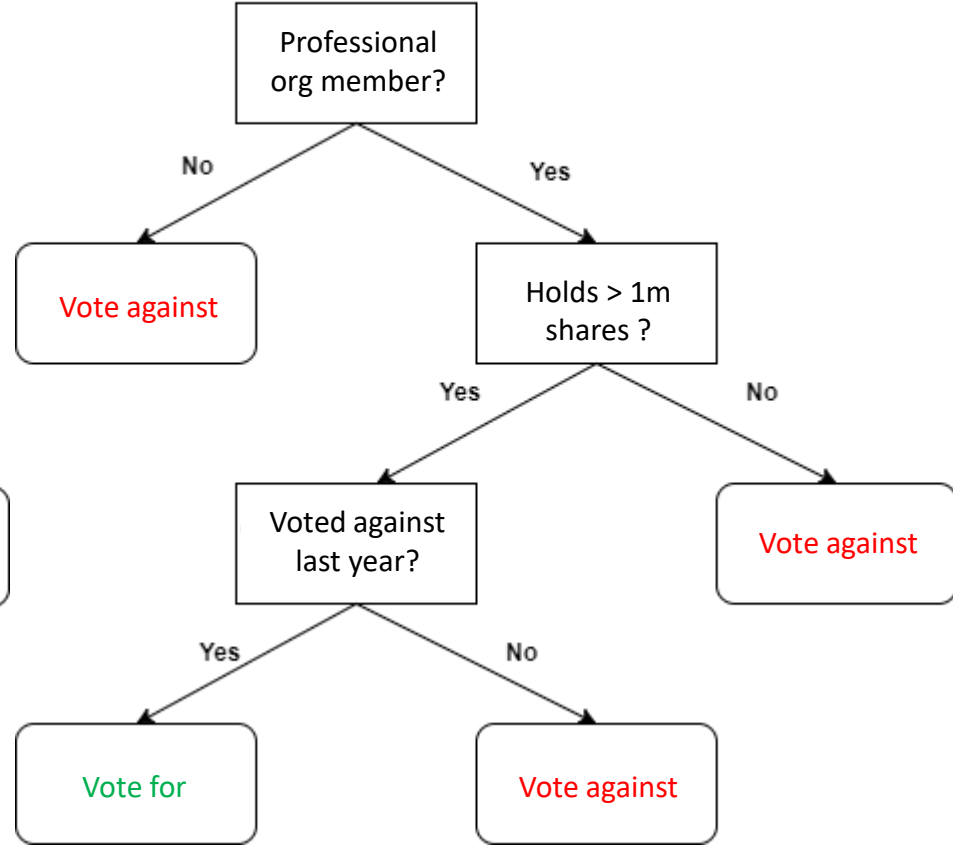
Results

PREDICTING SHAREHOLDER VOTING

Decision Tree for Classification and Prediction

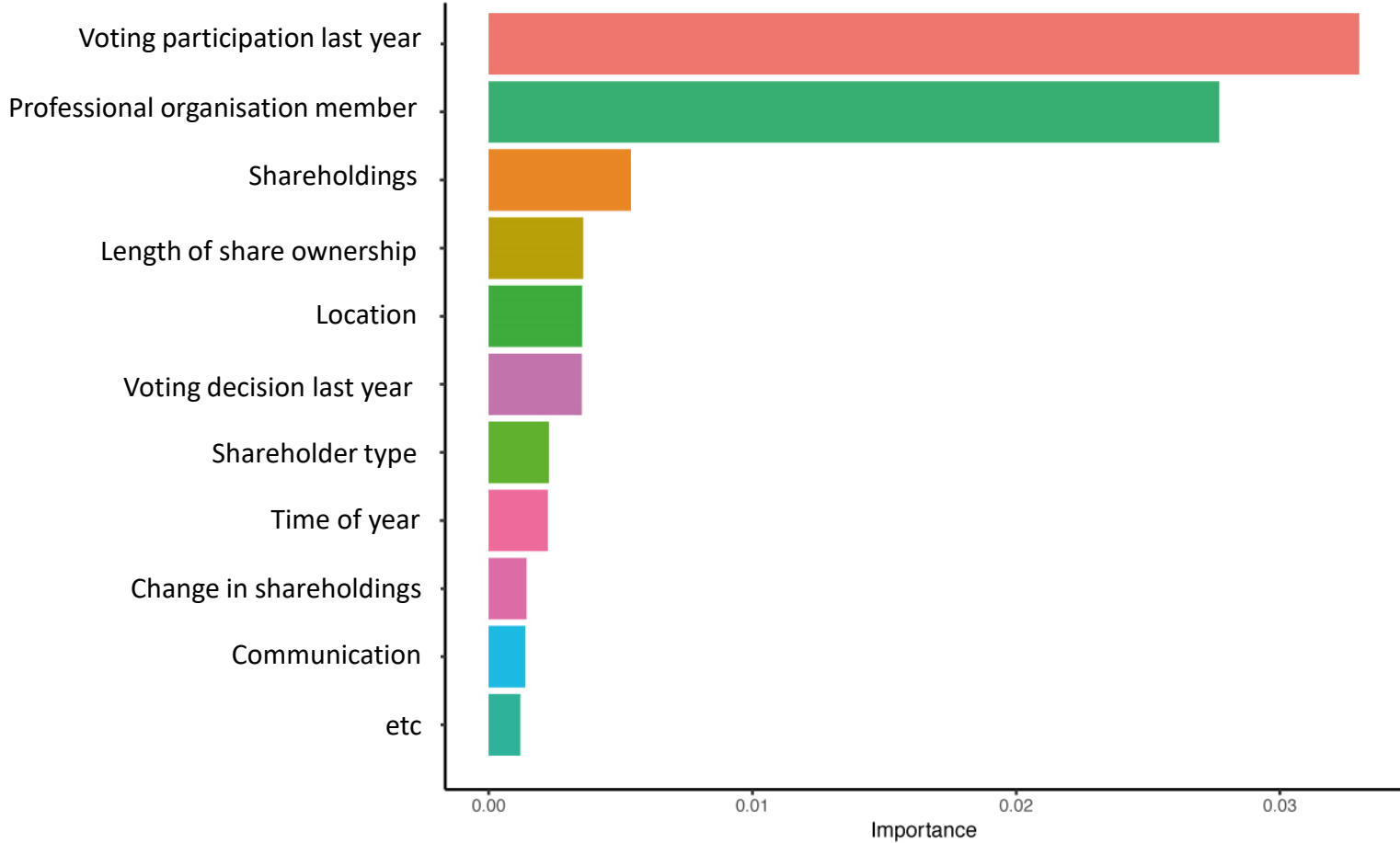


Decision Tree for Classification and Prediction



PREDICTING SHAREHOLDER VOTING

Overall study population




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keep.it.simple.stupid.



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HELEN TANNER

 0117 25 10100

 hello@data-cubed.co.uk

 data-cubed.co.uk