

Show rate analysis of Brazilian patients



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MEET OUR TEAM

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Deep Learning Spacial
Cardiologist

Luis Farre

Tiene un ordenador
GRIS SPECIAL



Deep Fake
Radiologist

Gabriela Pardo de Andrade

Se inventa la mitad de las cosas



Doctor Big Boss Magical
Deep Artificial Intelligence Master Mind

Daniel Eiroa



Universal global ML
analyst specialist scientist

Mar Lizana

Su ordenador funciona DPM



Machine Learning
Nutritionist

Dinis Oliveira

Se fue a portugal en medio
confinamiento

OUR TIMELINE

.....



CONTEXT

Based on a dataset of medical appointments with a 80/20 ratio show/no-show rate.



OBJECTIVE

"Predict the Show/No-Show rate"



STEPS

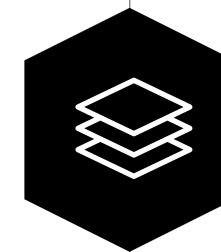
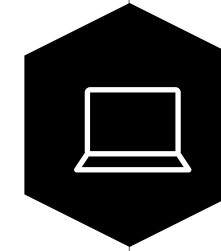
Analyze the no-show
Choose the right model to train and test

CHALLENGE

Sizable Data Set:

15 columns, 110.527 rows

Total of ~3 milion cells.



SOLUTION

Reduction of dimensionality:

Backward Elimination

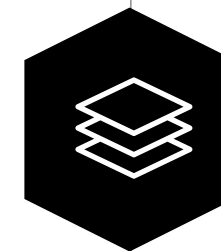
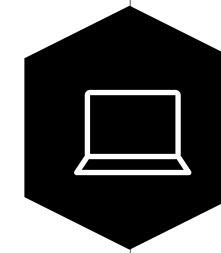
Feature Selection Random Forest Classifier

CHALLENGE

Unbalanced Data:

Show: 80%

No-show: 20%



SOLUTION

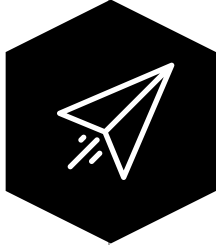
Undersampling:

Nearmiss

Random undersampling

CLASS DISTRIBUTION RESULTS

NEARMISS - 3
Show: 50%
No-show: 50%



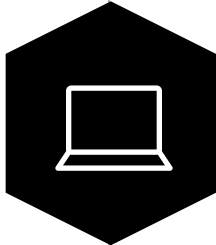
MODELS

SVC
RANDOM FOREST

| | model | accuracy | precision | recall | f1score | rocauc |
|---|--------------|----------|-----------|----------|----------|----------|
| 0 | logit | 0.635441 | 0.627157 | 0.668934 | 0.647243 | 0.689121 |
| 0 | knn | 0.630988 | 0.624124 | 0.647528 | 0.636971 | 0.674438 |
| 0 | decisiontree | 0.609940 | 0.611646 | 0.607589 | 0.609585 | 0.611078 |
| 0 | randomforest | 0.663403 | 0.643264 | 0.736862 | 0.686809 | 0.724306 |

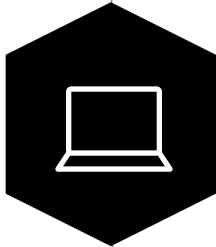
GRID SEARCH

Recall



METRICS

RECALL vs. ACCURACY



CONFUSION MATRIX

| <u>RF</u> | Positive | Negative |
|-----------|----------|----------|
| Positive | 1920 | 1460 |
| Negative | 405 | 2995 |

| <u>SVC</u> | Positive | Negative |
|------------|----------|----------|
| Positive | 2113 | 1267 |
| Negative | 502 | 2898 |



RESULTS

| | model | accuracy | precision | recall | f1score | rocauc | logloss |
|---|--------------|----------|-----------|----------|----------|----------|----------|
| 0 | randomforest | 0.720866 | 0.671878 | 0.856287 | 0.754818 | 0.810241 | 9.500883 |

| | | | | | | | |
|---|-----|----------|----------|----------|----------|----------|----------|
| 1 | svc | 0.726988 | 0.677477 | 0.863300 | 0.758651 | 0.790388 | 9.011815 |
|---|-----|----------|----------|----------|----------|----------|----------|

CONCLUSIONS:

A higher **RECALL** is preferable given that we prefer a **False Negative**

IMPROVEMENTS:

**Not done for computational reasons

ONEHOT ENCODING should be done for age and neighbourhood

Train with **SVC** model

Apply **DEEP LEARNING**



QUESTIONS?

THANK YOU



FMNDB CONSULTING
AI & ML Solutions

