



IDENTIFYING GLASS TYPES

Celia Ortega Vicente

WORKFLOW



```
graph LR; A[DATA COLLECTION] --> B[DATA PREPROCESSING]; B --> C[MODEL TRAINING];
```

DATA

COLLECTION

We collected glass samples from UCI Machine Learning Repository

DATA

PREPROCESSING

Clean the data to ensure its accuracy and ease of use.

MODEL

TRAINING

Test three models and to see which one is the most effective.

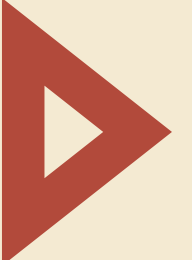
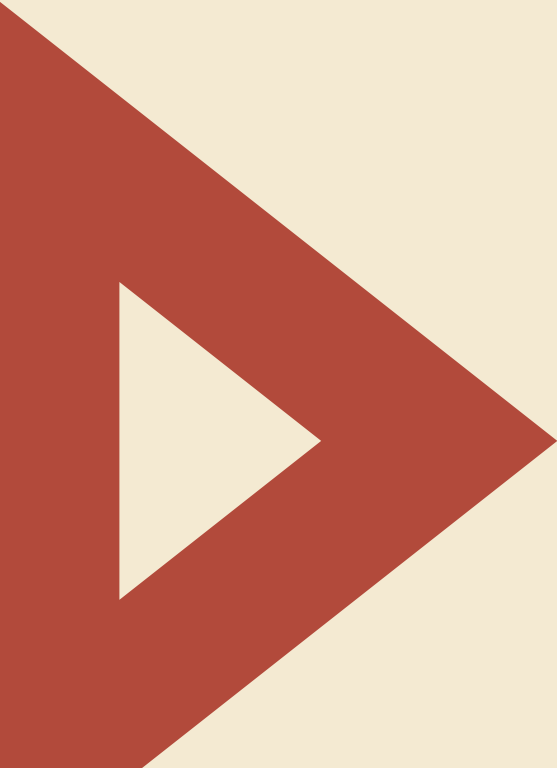
What for?

- Understanding the composition of glass samples.
- Could be crucial for criminal investigations.
- Investigating new methods for glass production to reduce cost and increase efficiency.





FEATURES

- **Refractive Index**
 - **Sodium (Na)**
 - **Magnesium (Mg)**
 - **Aluminum (Al)**
 - **Silicon (Si)**
 - **Potassium (K)**
 - **Calcium (Ca)**
 - **Barium (Ba)**
 - **Iron (Fe)**
- 
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Glass Types

BUILDING WINDOWS - FLOAT PROCESSED

BUILDING WINDOWS - NON-FLOAT PROCESSED

VEHICLE WINDOWS - FLOAT PROCESSED

CONTAINERS

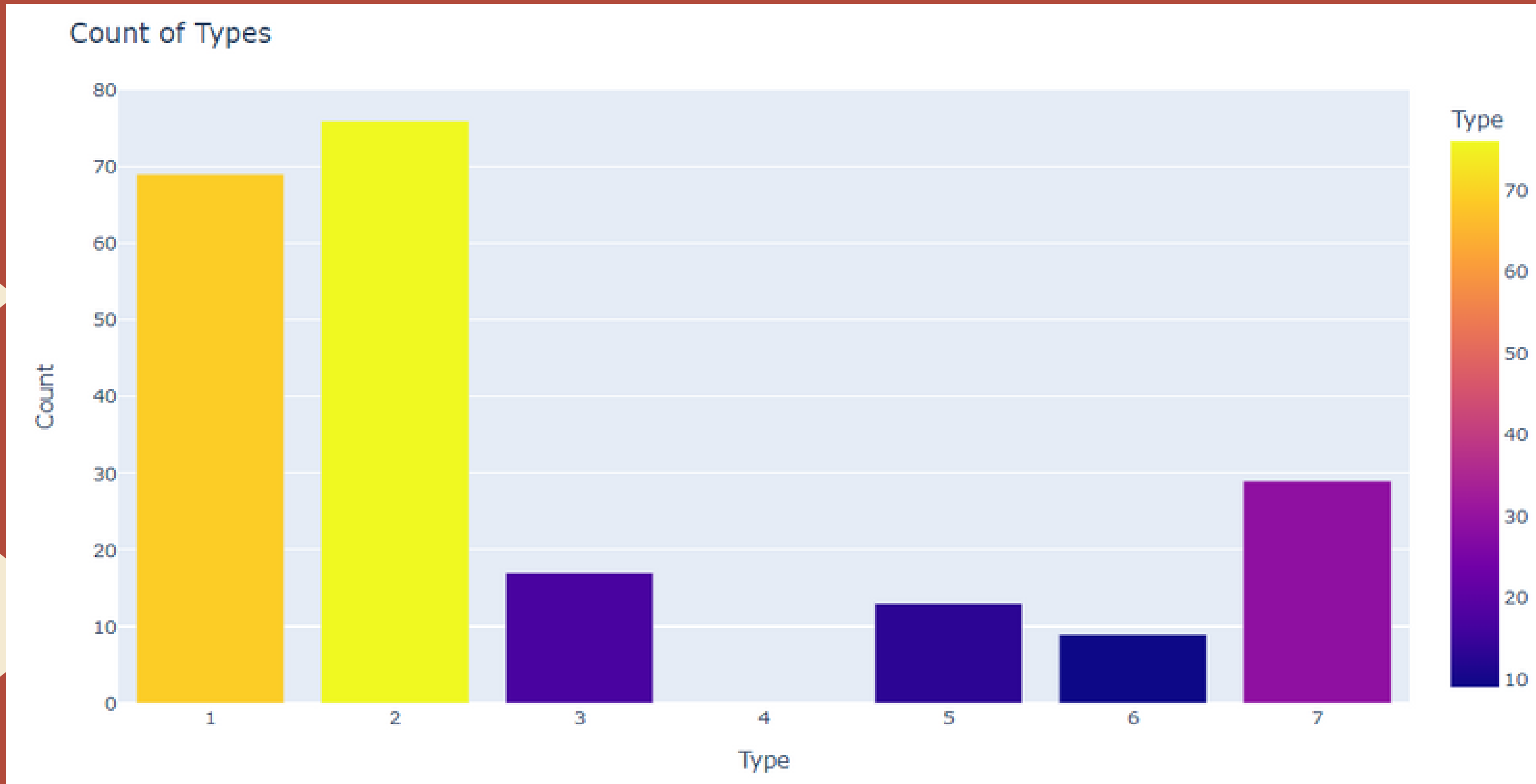
▶ TABLEWARE

HEADLAMPS



DATA IMBALANCE PROBLEMS

- Missing one type of glass in data (type 4)
- Large difference in sample sizes between classes



MODELS

1

Random Forest

Kappa score

- With imbalanced data: 0.785
- With balanced data: 0.756

2

Gradient Boosting Classifier

Kappa score

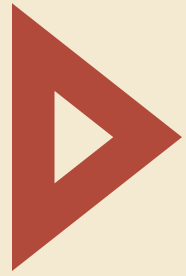
- With imbalanced data: 0.714
- With balanced data: 0.658

3

K-Nearest Neighbours

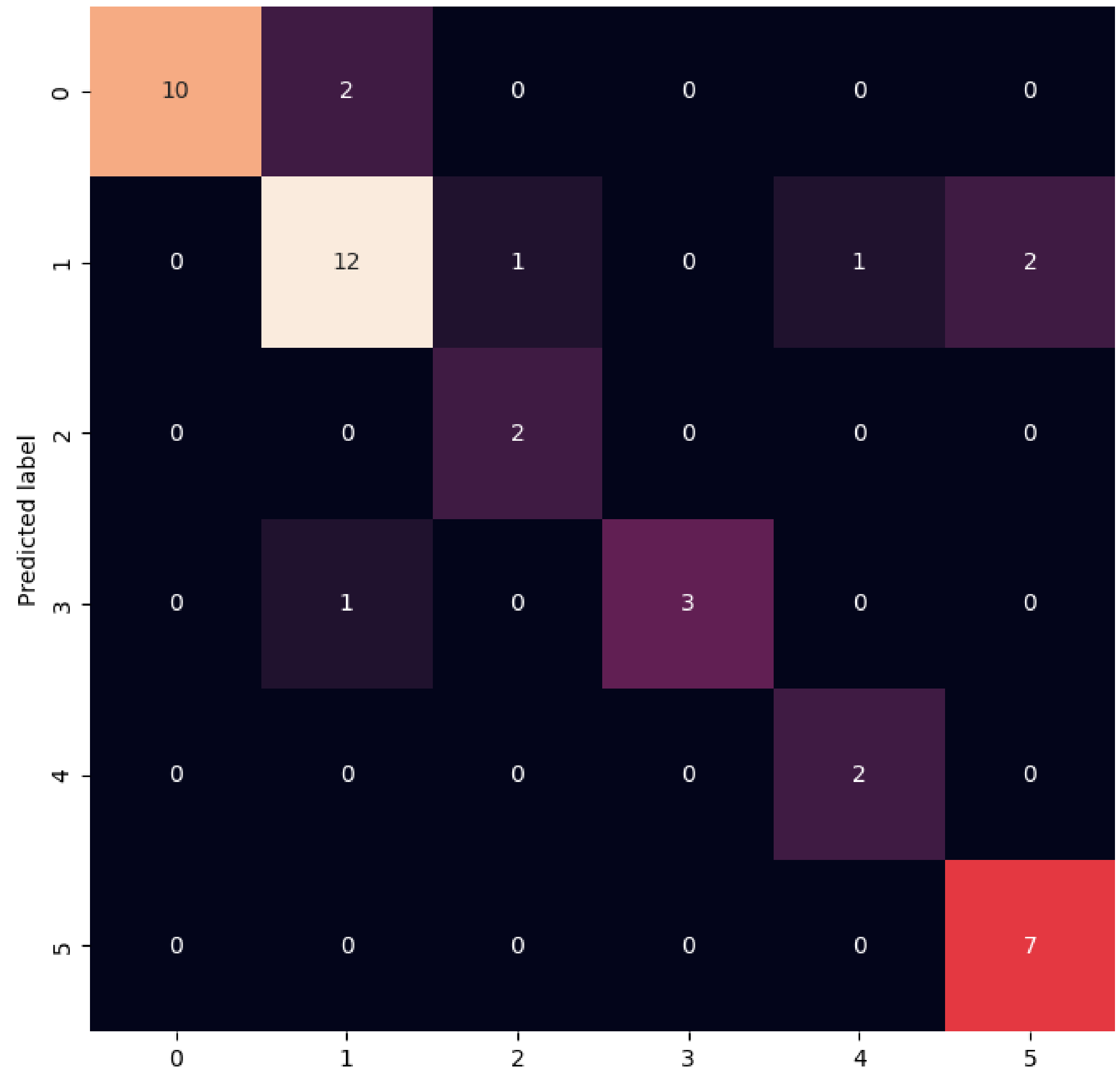
Kappa score

- With imbalanced data: 0.534
- With balanced data: 0.664



CONFUSSION MATRIX

ONLY THE 16% OF
THE DATA WON'T BE
GOOD PREDICTED



**TRY IT
YOURSELF**

