# Nkululeko: Machine Learning Experiments on Speaker Characteristics Without Programming

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# Introduction & Summary

- A tool to do machine learning without the need to program
- Focused on speech, but works generally on audio
- Opensource project, 100% Python: github.com/felixbur/nkululeko
- Targets combinations of feature sets and machine learners
- Uses configuration file templates

#### Example config

[EXP] root = ./

## Learners and features

#### **Features**

Expert features: opensmile, Praat Embeddings: Wave2vec2.0, TRILL, CLAP

Learners SVM, SVR, K-nn, Tree, XGB, XGR, GMM, MLP

# **Result presentation**

The main format of visualizing Nkululeko experiment outcomes is the

```
name = exp_emodb
[DATA]
databases = ['emodb']
emodb = ./emodb/
emodb.split_strategy = speaker_split
emodb.test_split = 30
target = emotion
labels = ['anger', 'boredom', 'neutral', 'sadness']
[FEATS]
type = ['os'] # opensmile features
[MODEL]
type = xgb # xg-boost classifier
```

Nkululeko is organized in modules:

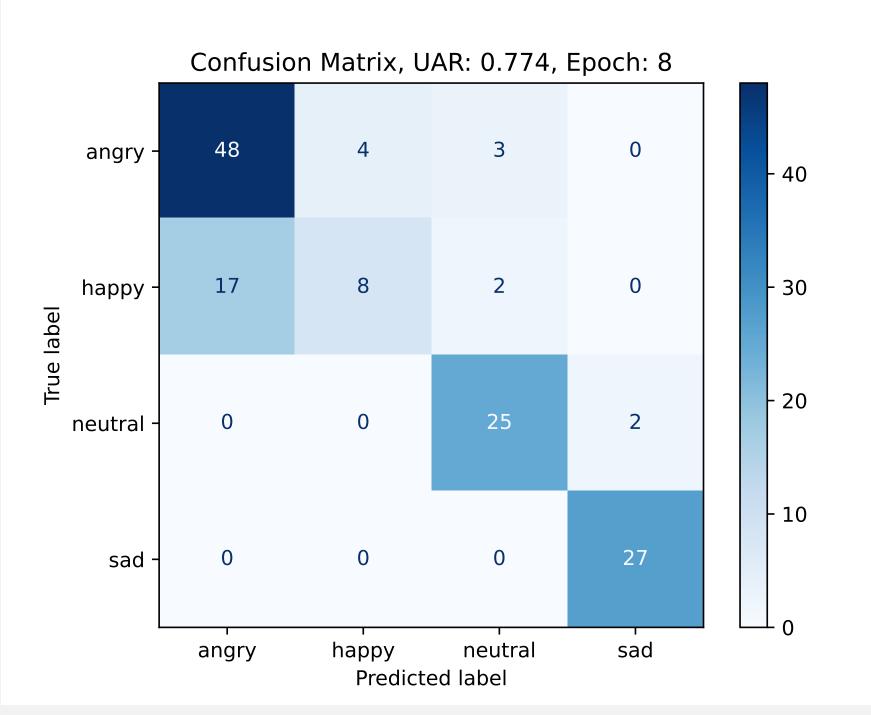
- explore: data exploration
- nkululeko: machine learning
- augment: data augmentation
- demo: model demoing

# Motivation

- Empiricists sometimes struggle with programming, teaching students
- Getting quick overview on databases

confusion matrix.

#### Confusion matrix



#### Figure: Confusion matrix per emotion

#### Epoch progression

# Data exploration

#### Sample distribution

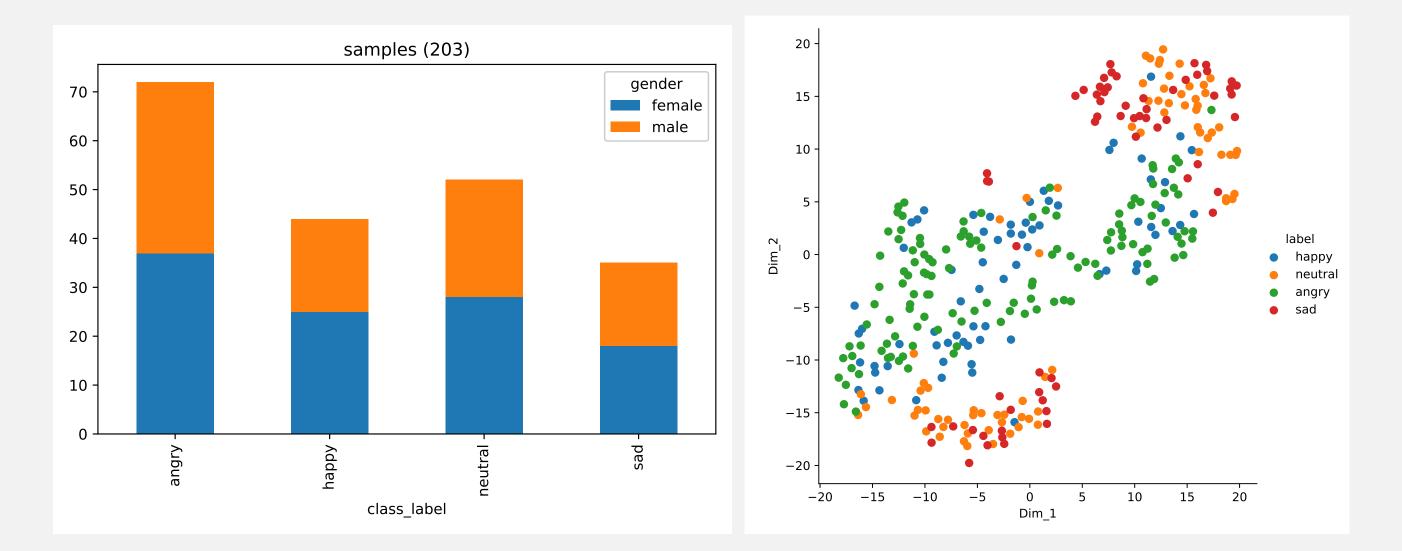
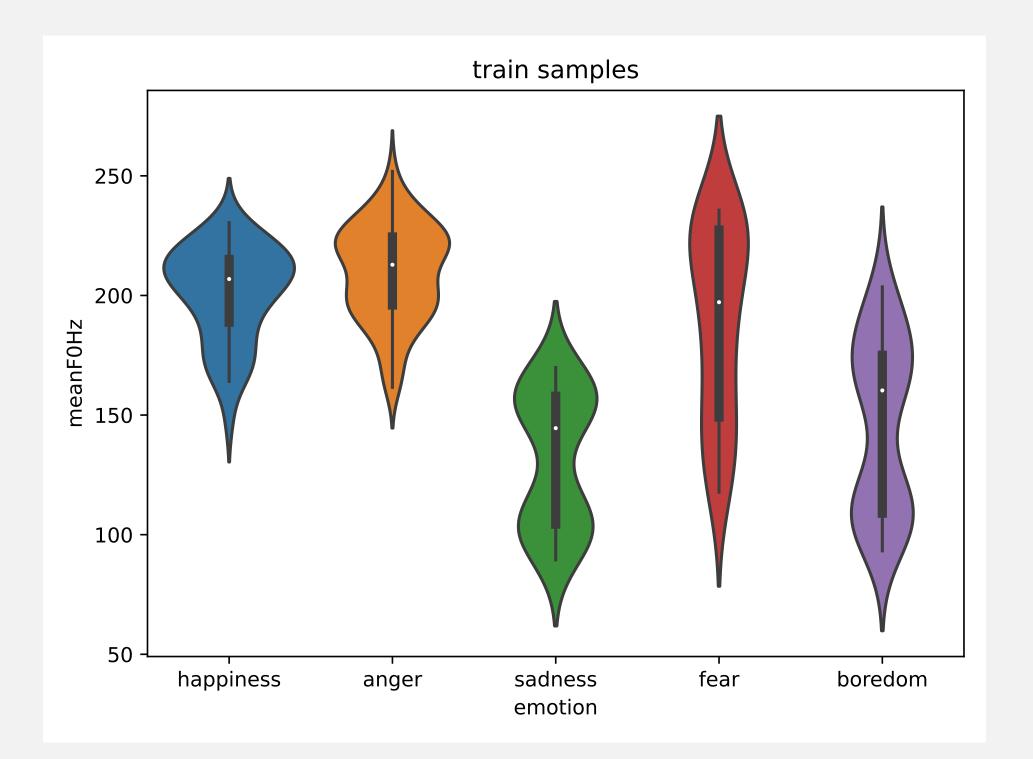


Figure: Emotion distribution per gender and t-SNE plot of features

#### Feature exploration



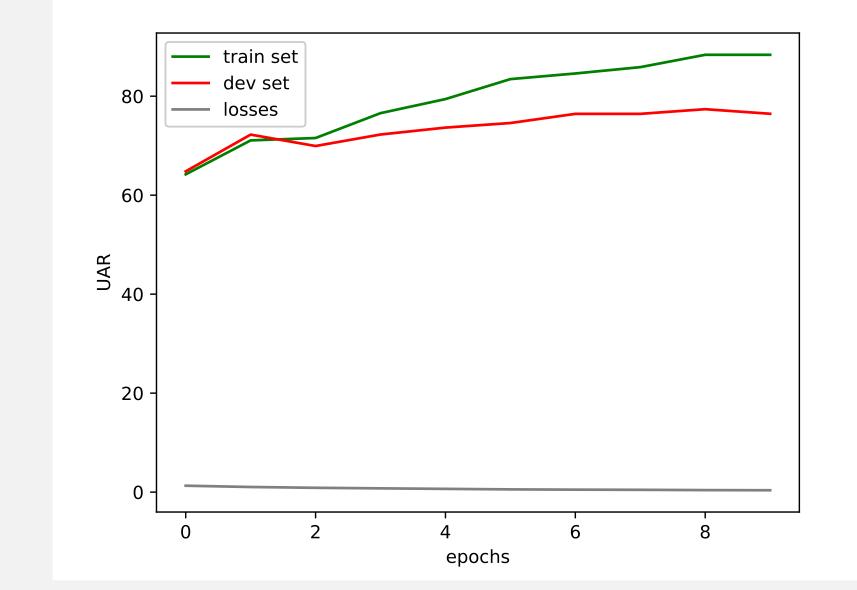


Figure: Progression of unweighted average recall per epoch

# Data import

Data can be imported with a simple csv format: file, speaker, age, emotion f1.wav, mark, 34, happy .... f123.wav, lisa, 21, sad

Figure: Violinplots for selected features

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