



LEVERAGING TINYML FOR ACOUSTIC-BASED POULTRY HEALTH MANAGEMENT

Research Talk Presented at the Workshop on widening access to TinyML Network by Establishing Best Practices in Education held at ICTP, Trieste, Italy (July 3 – 7 2023)



This is Bowen University!

- A 22-year-old foremost private University in Nigeria
- Located on a hill
- About 4 hours drive from Lagos
- Average population of 5000 students





Some of our Previous activities

- Leveraging Radio Telemetry and Data Analytics to track Fruit Bat Movement Pattern ([Data Science Africa Research Grant](#))
- K'eyemaba: Pest Detection and Prevention using Unmanned Aerial Vehicle on Rice Farmland
- CNN-Based Crop disease detection on Cassava Farm
- Organised TinyML workshop bringing some students from four different Universities together for two days

*Our works and challenges are geared towards providing solution to local challenges which can be scaled for global usefulness— **Glocalization of ideas!***



THE BASIS



Increase in consumption of animal products such as meat, milk and eggs in Nigeria but decline in production of these sources of protein

Growing concern of increasing food production by 25% to meet the needs of the growing population

Major source of protein – poultry farms are faced with health challenges which leads to economic loss in poultry farm

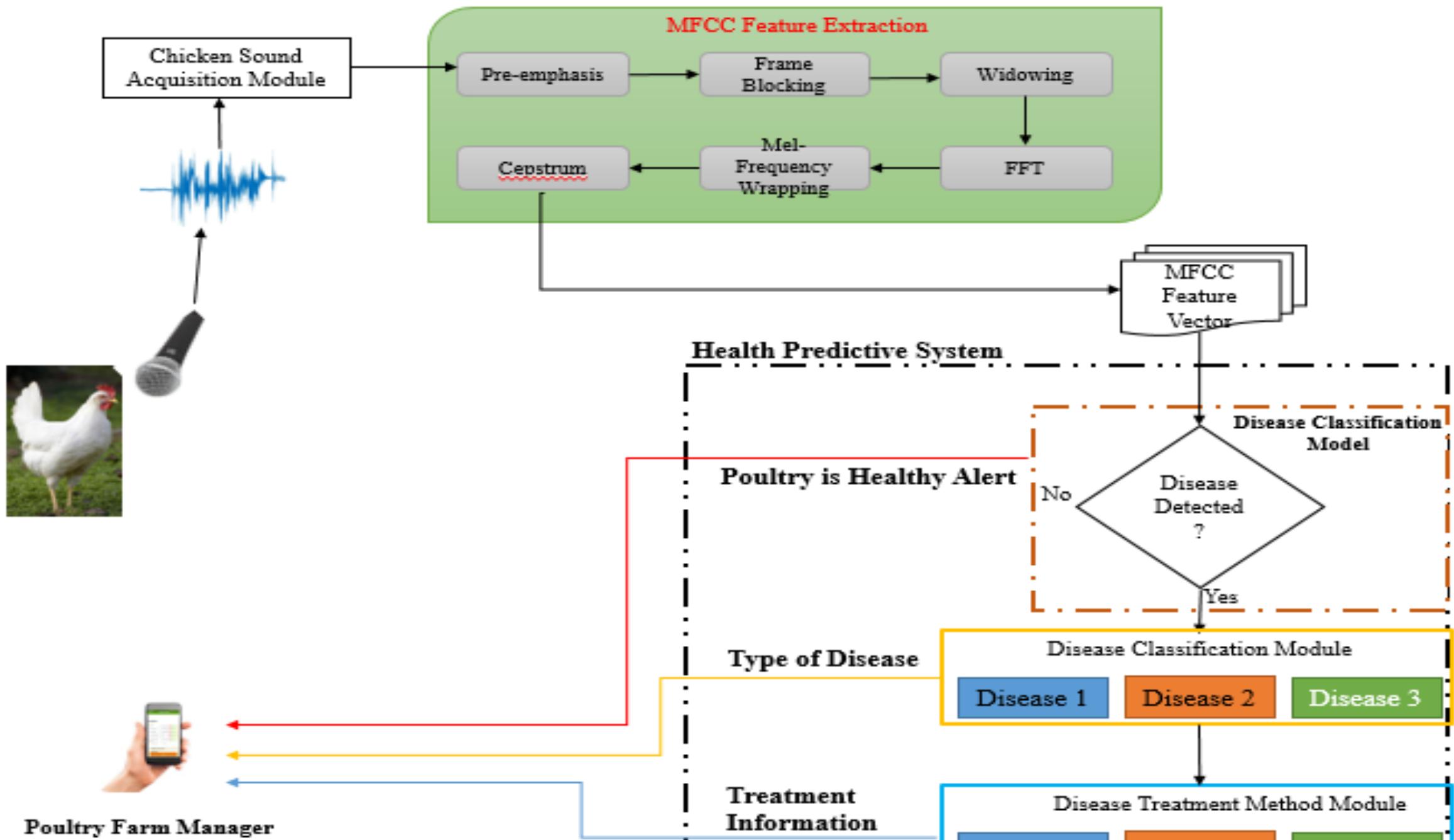
Hence, the need for early detection and prevention of poultry related health challenges

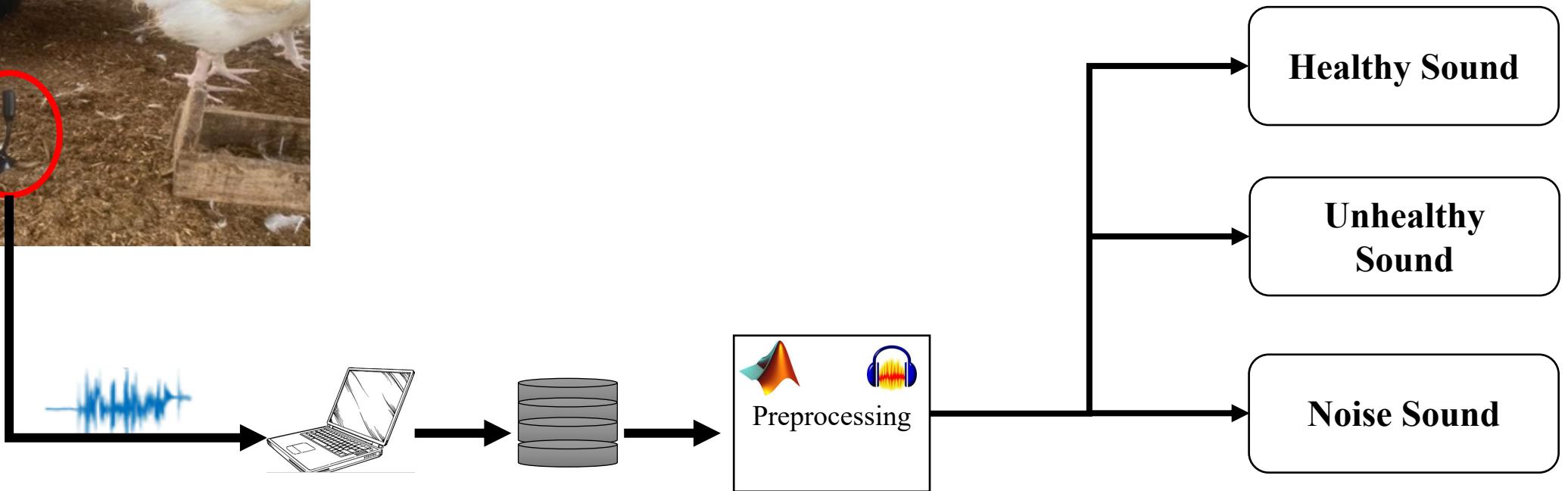
OUR FOCUS

Paucity of Publicly available datasets
on Environmental Sound
Classification

Most studies so far focus on specific
species of chicken with many of
these datasets not publicly available
in open-source repository such as
GitHub, Kaggle, etc.

The work is set out to achieve
gathering and annotation of large
dataset of poultry vocals and
acoustics under varying weather
conditions for the purpose of stress
detection and prevention to mitigate
loss and thereby, maximize profit on
poultry farms







DAY OLD

A total of 100 broiler chickens were collected and divided into two groups of 50 each. One of the group were treated for diseases while the other were not.

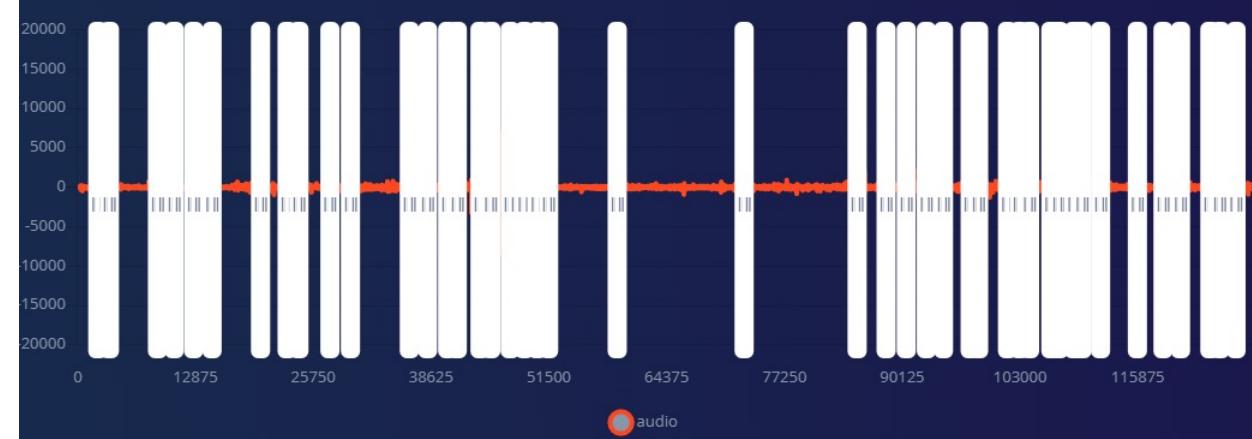
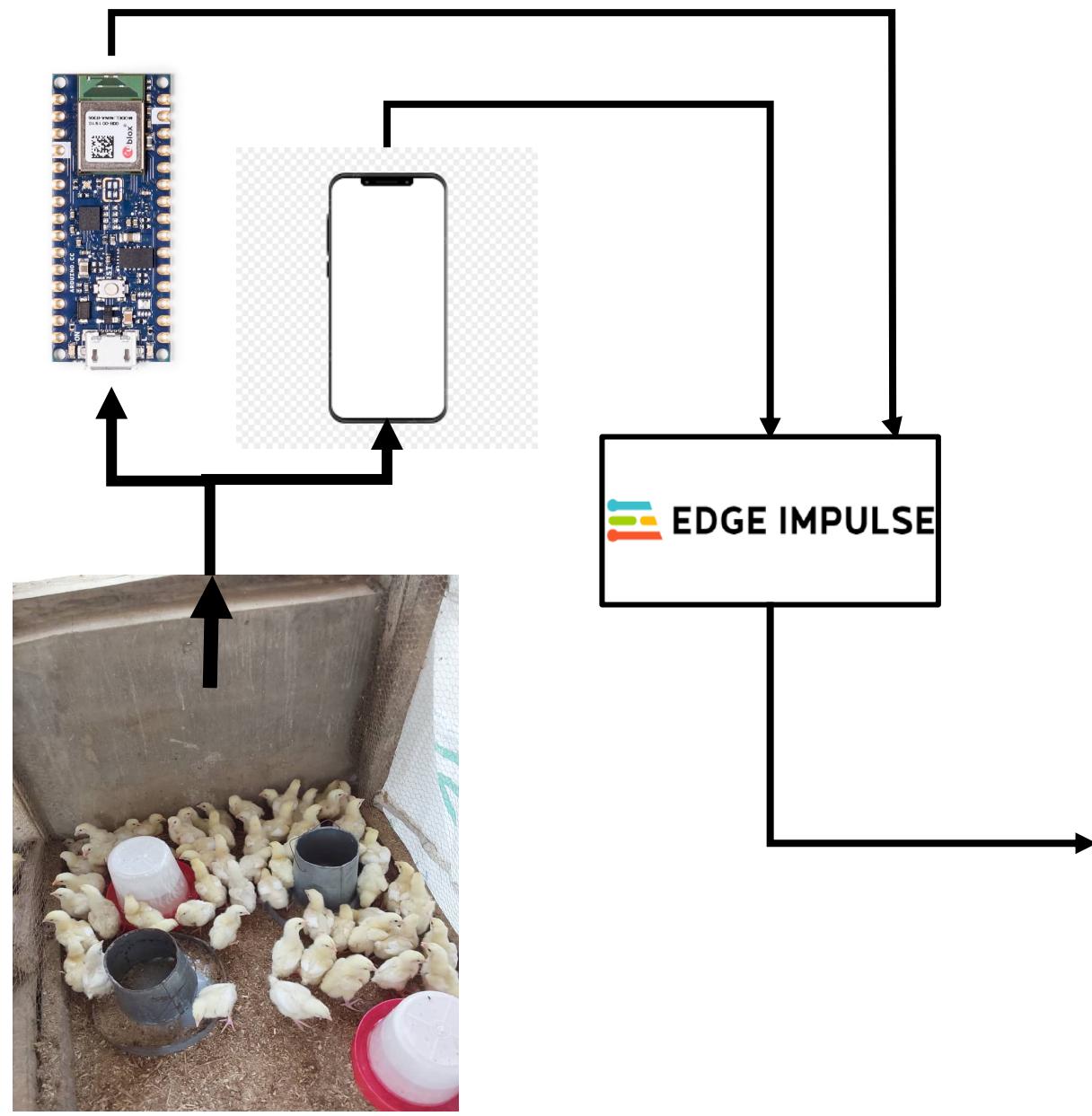


2 WEEKS OLD



8 WEEKS OLD

DATA COLLECTION WITH SENSORS



Screenshot of the Edge Impulse web interface, showing a dataset for "Segun Adebayo / Chicken Audio".

Left Sidebar: Includes links for Dashboard, Devices, Data acquisition, Impulse design (Create impulse, EON Tuner, Retrain model, Live classification, Model testing, Performance calibration, Versioning, Deployment), and Getting Started (Documentation, Forums).

Main Area:

- Dataset Summary:** Shows "DATA COLLECTED 47s" and "TRAIN / TEST SPLIT 0% / 100%".
- Dataset Table:** Lists 10 entries under "Training (0)" and "Test (47)".

SAMPLE NAME	LABEL	ADDED	LENGTH
Sick-bird-sound-1.s47	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
Sick-bird-sound-1.s46	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
Sick-bird-sound-1.s45	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
Sick-bird-sound-1.s44	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
Sick-bird-sound-1.s43	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
Sick-bird-sound-1.s42	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
Sick-bird-sound-1.s41	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
Sick-bird-sound-1.s40	Sick-bird-sound-1	Apr 17 2023, 14:59:15	1s
- Raw Data Plot:** A spectrogram titled "Sick-bird-sound-1.s46" showing amplitude over time and frequency. A red circle labeled "audio" is at the bottom right.
- Control Panel:** Includes buttons for "Collect data" and "Connect a device to start building your dataset".

DATASET DESCRIPTION

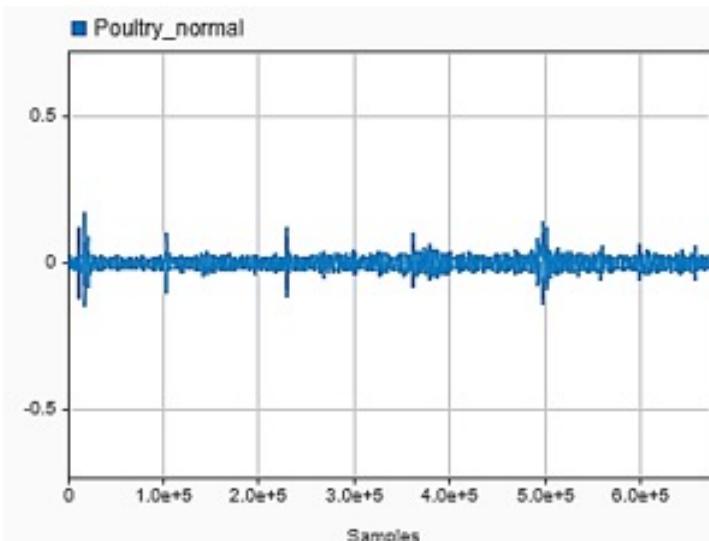
The dataset contains 346 audio signal files and grouped into three folders: healthy, noisy, and unhealthy.

There are 139 audio files in the healthy folder, 86 in the noise folder, and 121 in the unhealthy folder.

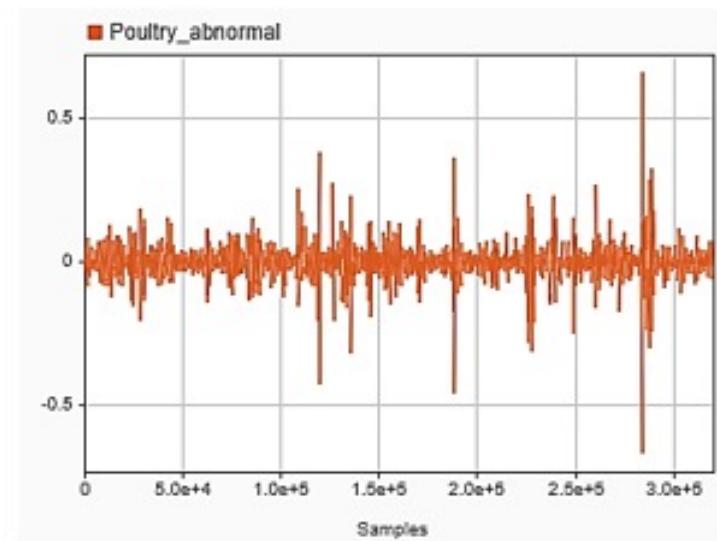
The audio signal file have a range of 5 seconds to 60 seconds length.

The selected sound segments in the unhealthy folder include chicken cough, snore and rale sound.

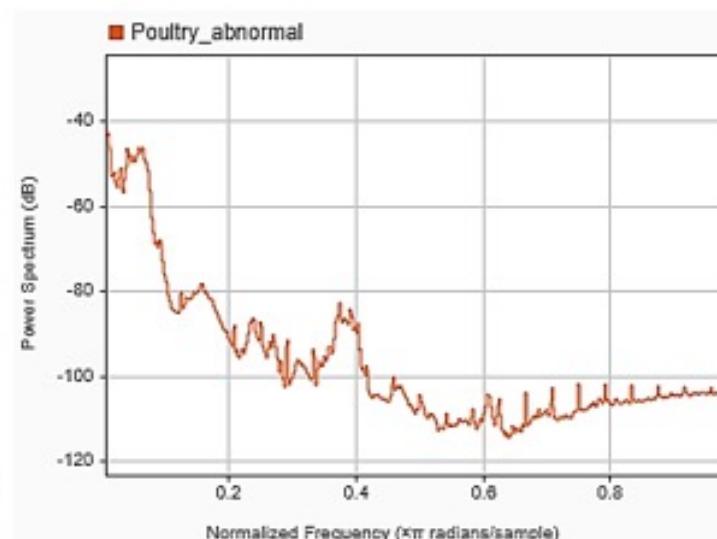
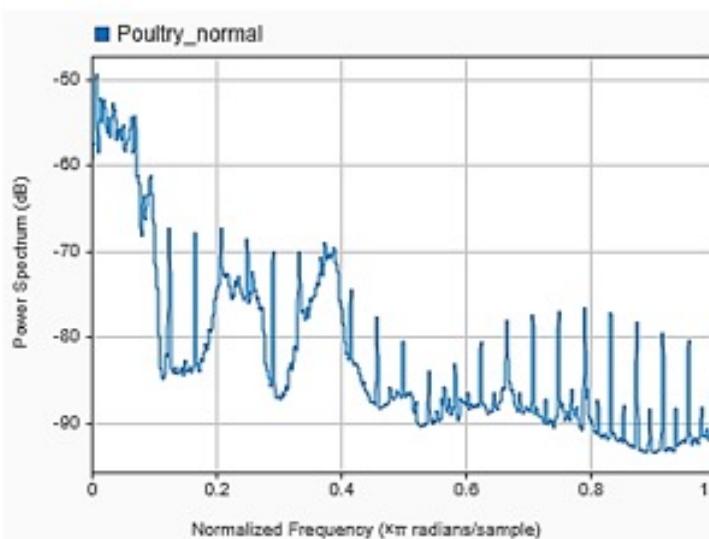
The Noise folder contain sound segments that include background noises (e.g., moving vehicle and human voices) and sound created by poultry bird activities (e.g., feeding, pecking one another).



(a)



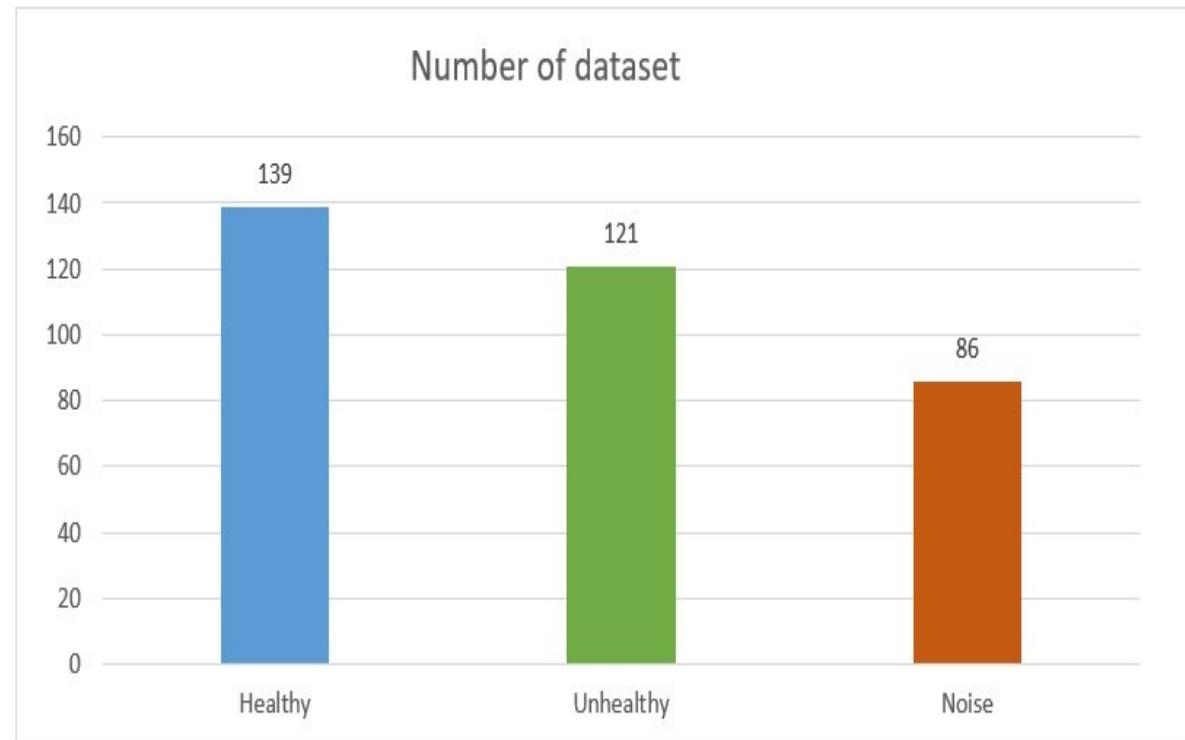
(b)



Dataset Description

A total of 346 audio signal files were successfully collected from the poultry farm. This dataset comprises of

- 139 files that represent healthy audio signals,
- 121 files represent unhealthy audio signals, and
- 86 files are the noise audio signals.



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Poultry Vocalization Signal Dataset for Early Disease Detection

Published: 12 June 2023 | Version 1 | DOI: [10.17632/zp4nf2dxbh.1](https://doi.org/10.17632/zp4nf2dxbh.1)

Contributors: [Halleluyah Aworinde](#), [Segun Adebayo](#), [Akinwale Akinwunmi](#), [Olufemi Alabi](#), [Adebamiji Ayandiji](#), [Olaide Oke](#), [Abel Oyebamiji](#), [Adetoye Adeyemo](#), [Aderonke Sakpere](#), [Kizito Echetama](#)

Description

The aim of generating this dataset is to deliver open, accessible and quality machine learning dataset for Poultry farm management. The dataset contains a total of 346 audio signal files which are divided into three folders: healthy, noise and unhealthy. The healthy folder contains 139 audio files, noise folder has 86, while unhealthy folder has 121 files. The audio files are saved in .wav file format.

[Download All 869 MB](#)

Files

Chicken_Audio_Dataset

Steps to reproduce

Bowen University poultry research farm was used as the experimental environment where total number of day-old chicks were acquired and divided into two groups. The first group was treated for respiratory diseases and the other group was not treated for respiratory diseases. The chicks were then placed in a controlled environment, separately from each other, and microphones were positioned at a suitable distance from the birds to avoid background noise and other sounds that may interfere with the analysis. Data was collected at 96 kHz with 24-bit samples. The audio data collection process occurred continuously, three times daily (morning, afternoon and night) for a period of 65 days. During this period, food and water are readily available to the birds at all times. After 30 days, the untreated group developed respiratory disease which was evident in their sound. These data were also captured as unhealthy. Audio signals of poultry chickens were collected and stored in MA4 and later converted to WAV format.

Dataset metrics

Usage

Downloads:

693

Views:

22

[View details >](#)

Latest version

Version 1

Published:

12 Jun 2023

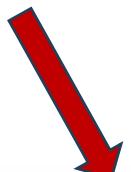
DOI:

[10.17632/zp4nf2dxbh.1](https://doi.org/10.17632/zp4nf2dxbh.1)

Cite this dataset

Aworinde, Halleluyah; Adebayo, Segun; Akinwunmi, Akinwale; Alabi, Olufemi; Ayandiji, Adebamiji; Oke, Olaide; Oyebamiji, Abel; Adeyemo, Adetoye; Sakpere, Aderonke; Echetama, Kizito (2023), "Poultry Vocalization Signal Dataset for Early Disease Detection", Mendeley Data, V1, doi: [10.17632/zp4nf2dxbh.1](https://doi.org/10.17632/zp4nf2dxbh.1)

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Poultry Vocalization Signal Dataset for Early Disease Detection

Publication Year: 2023

715

Usage

Home

Overview

Highlights

Metrics Details

USAGE	715
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Mendeley Data ↗	693
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Bibliographic Details

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URL ID: <https://data.mendeley.com/datasets/zp4nf2dxbh>

↗; <http://dx.doi.org/10.17632/zp4nf2dxbh.1> ↗;

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<https://data.mendeley.com/datasets/zp4nf2dxbh/1> ↗

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Aderonke Echituzo, Kizito

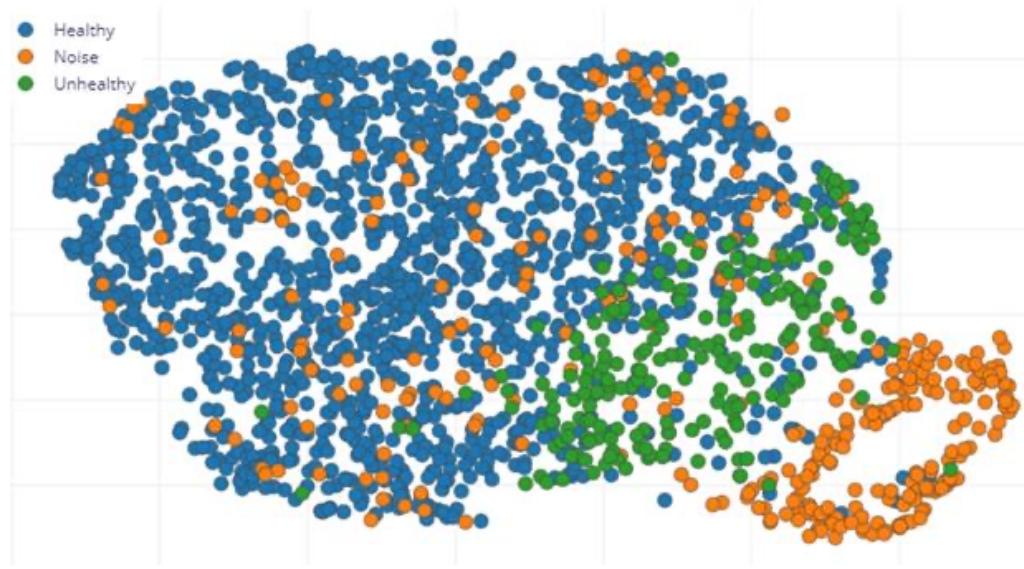
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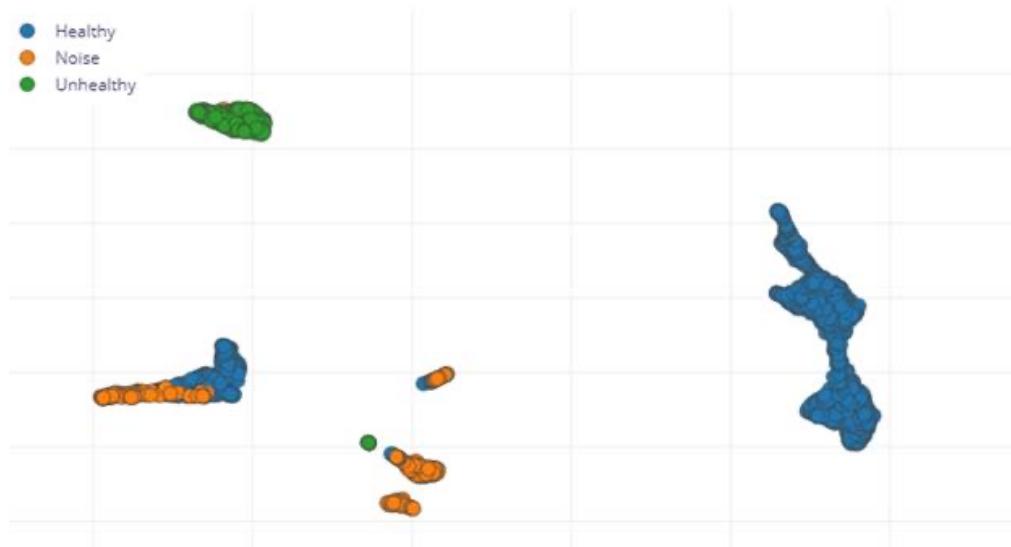
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INSIGHTS FROM DATASET FEATURES

MFCC Features



Spectrogram Features



MFE Features



INSIGHTS FROM MODEL TRAINING

Effect of Data Argumentation of Trained Model

Add noise	None	Low	High	Low	High	High	High
Mask time bands	None	None	None	Low	High	High	Low
Mask frequency bands	None	None	None	Low	High	None	None
Accuracy	96.4%	87.8%	80.2%	72.8%	39.0%	60.1%	57.8%
F1-score	96.3%	87.3%	79.6%	70%	28.3%	58%	55.3%
Loss	0.14	0.36	0.46	0.69	1.27	0.87	0.89

Performance of Neural Network Training using MFCC Features

Last training performance (validation set)



ACCURACY
85.6%

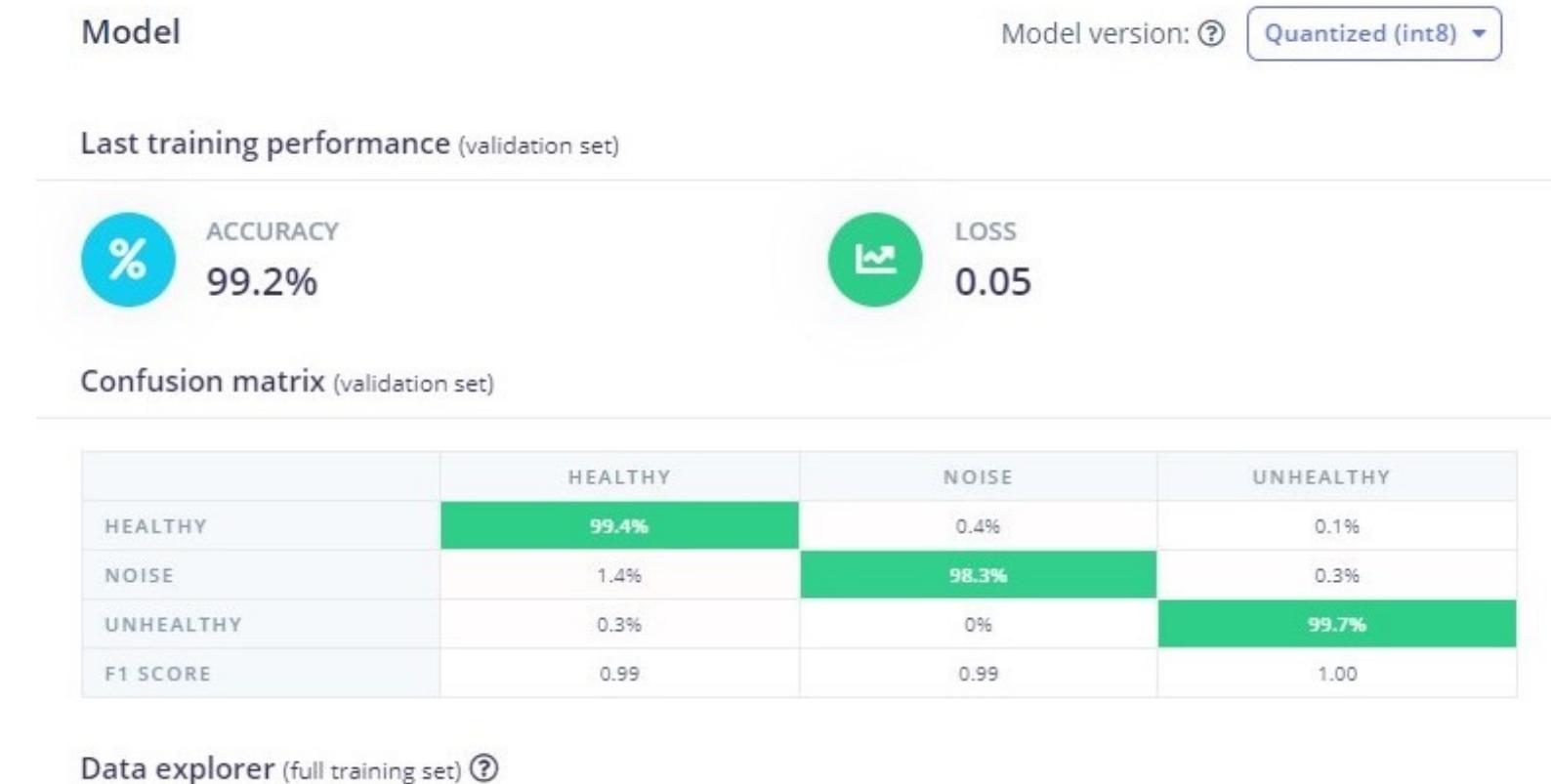


LOSS
0.38

Confusion matrix (validation set)

	HEALTHY	NOISE	UNHEALTHY
HEALTHY	91.3%	4.9%	3.8%
NOISE	23.3%	72.2%	4.5%
UNHEALTHY	8.5%	1.7%	89.8%
F1 SCORE	0.86	0.79	0.90

Performance of Neural Network Training using Spectrogram Features



Performance of Neural Network Training using MFE Features

Model

Model version: ②

Quantized (int8) ▾

Last training performance (validation set)



ACCURACY
97.4%



LOSS
0.08

Confusion matrix (validation set)

	HEALTHY	NOISE	UNHEALTHY
HEALTHY	99.7%	0.2%	0.1%
NOISE	8.7%	91.2%	0.1%
UNHEALTHY	0.4%	0%	99.6%
F1 SCORE	0.97	0.95	1.00

Performance of Neural Network Training using MFE & MFCC Features

Last training performance (validation set)



ACCURACY
96.8%



LOSS
0.11

Confusion matrix (validation set)

	HEALTHY	NOISE	UNHEALTHY
HEALTHY	97.4%	1.9%	0.7%
NOISE	6.1%	93.2%	0.7%
UNHEALTHY	0.5%	0.1%	99.4%
F1 SCORE	0.97	0.95	0.99

RESULTS

Types of Model	Model Accuracy	F1 Score	Loss
MFCC Features-based	85.6%	85%	0.38
Spectrogram Features-based	99.2%	99.3%	0.05
MFE Features-based	97.4%	97.3%	0.08
MFCC & Spectrogram Features-based	98.4%	98.6%	0.06
MFCC & MFE Features-based	96.8%	97%	0.11

Team Composition



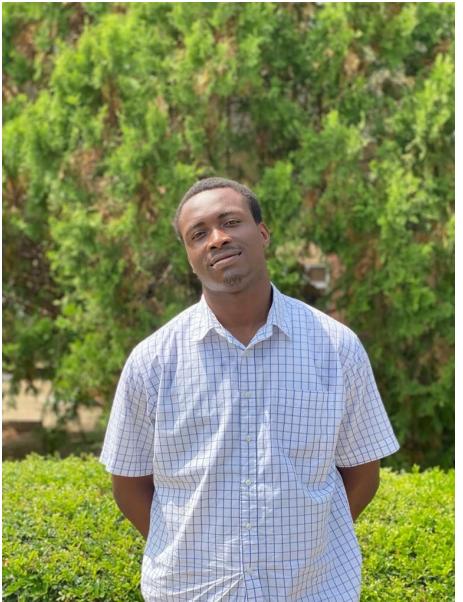
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Machine Intelligence



Dr. Akinwale Akinwunmi
Cloud Computing



Dr. Adebamiji Ayandiji
Agricultural & Rural Extension Specialist



Echetama Kizito Chizitere
Student



OKE OLAIDE FELICIA
Student



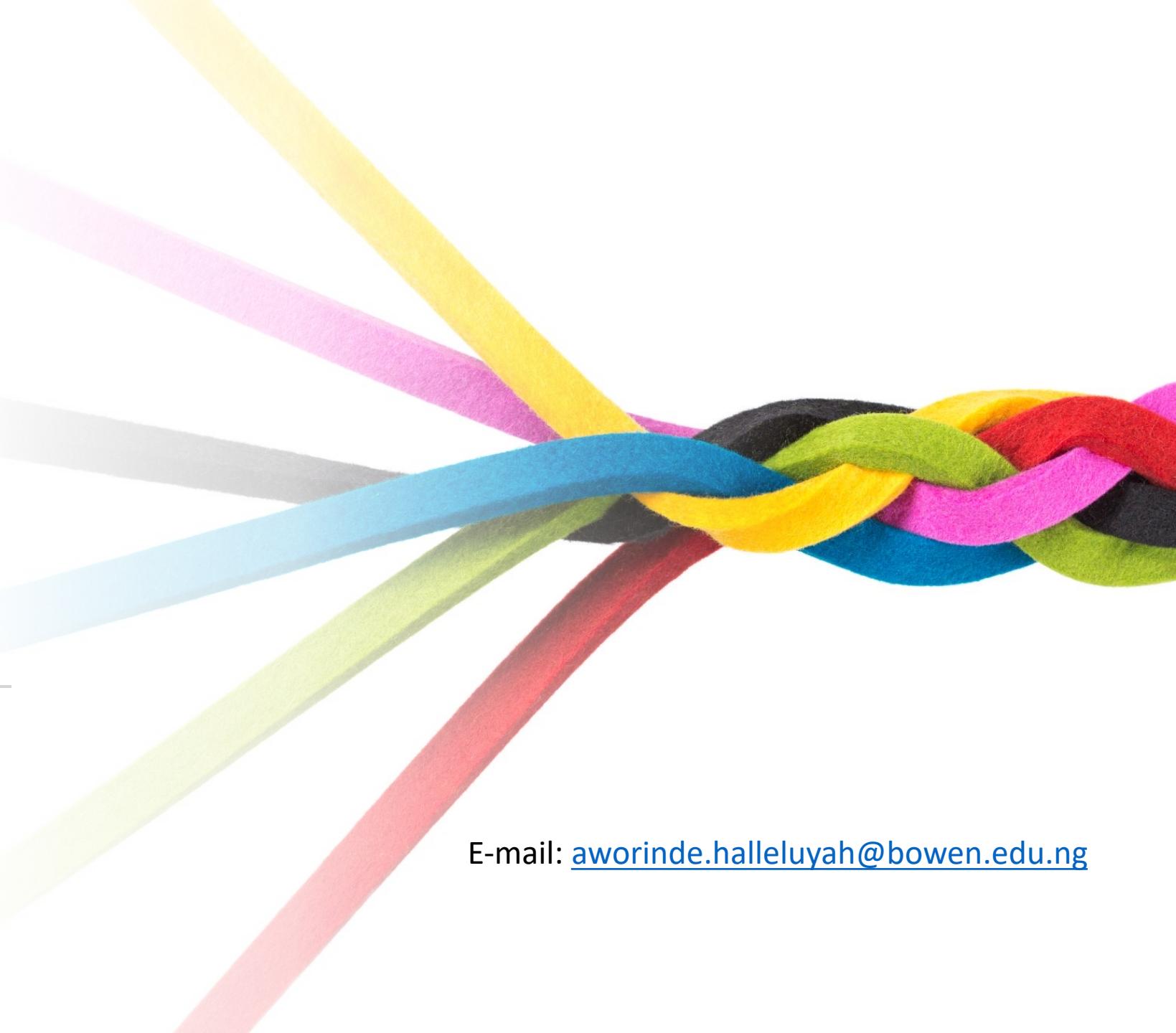
Dr. Segun Adebayo
IoT/Hardware



Prof. Olufemi Alabi
Animal Scientist



We are still
learning, we
open to support
and
collaboration



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Excellence & Godliness



THANK YOU...