



# 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 4hours 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



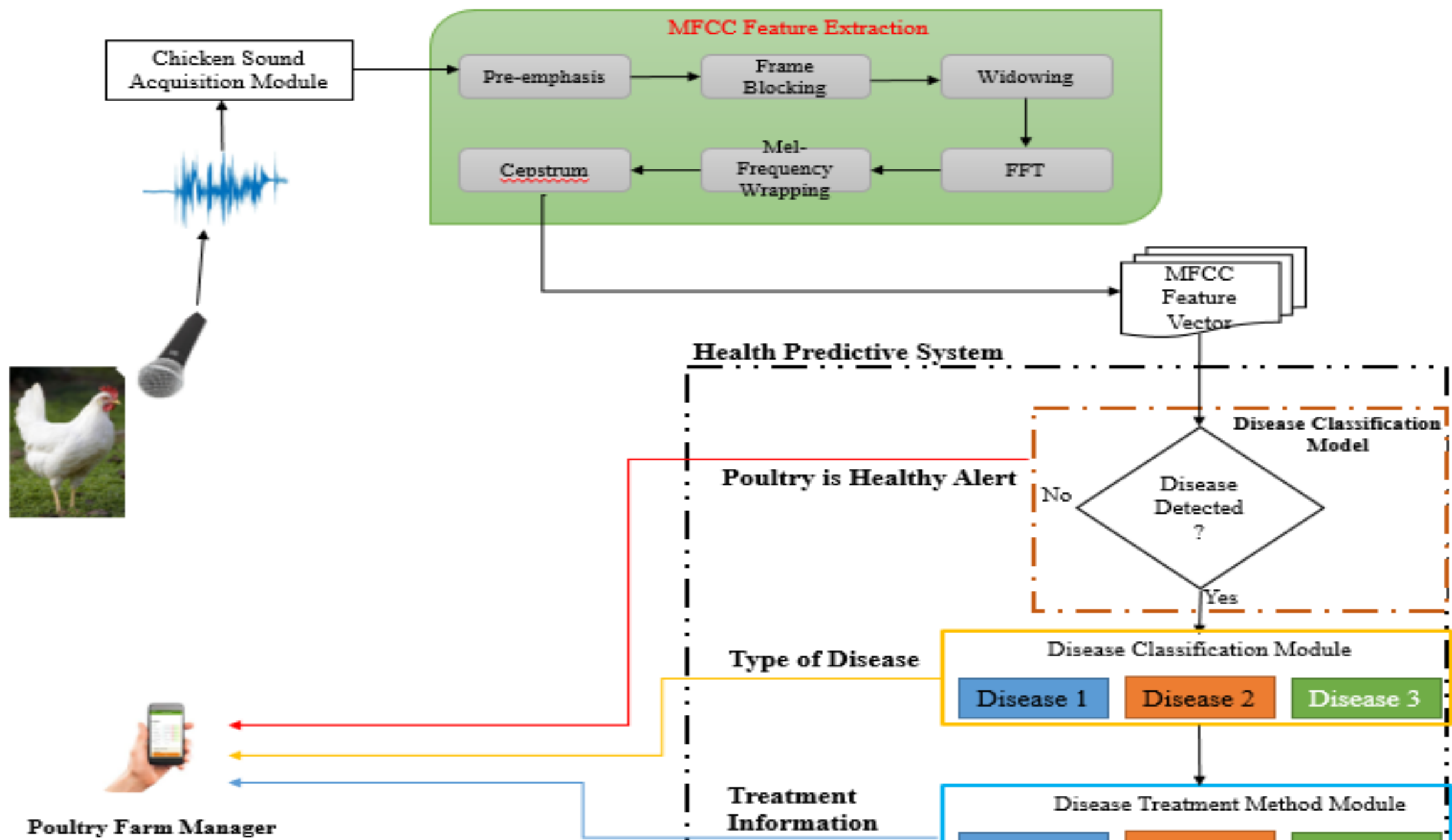
# OUR FOCUS

Paucity of Publicly available datasets  
on Environmental Sound  
Classification

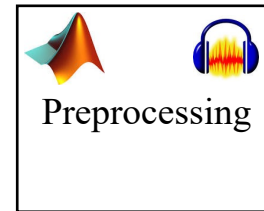
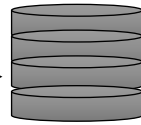
Most studies so far focus on specific  
species of chicken with many of  
these datasets not publicly available  
in open-source repository such as  
GitHub, Kaggle, etc.

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









**Healthy Sound**

**Unhealthy  
Sound**

**Noise Sound**





**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.



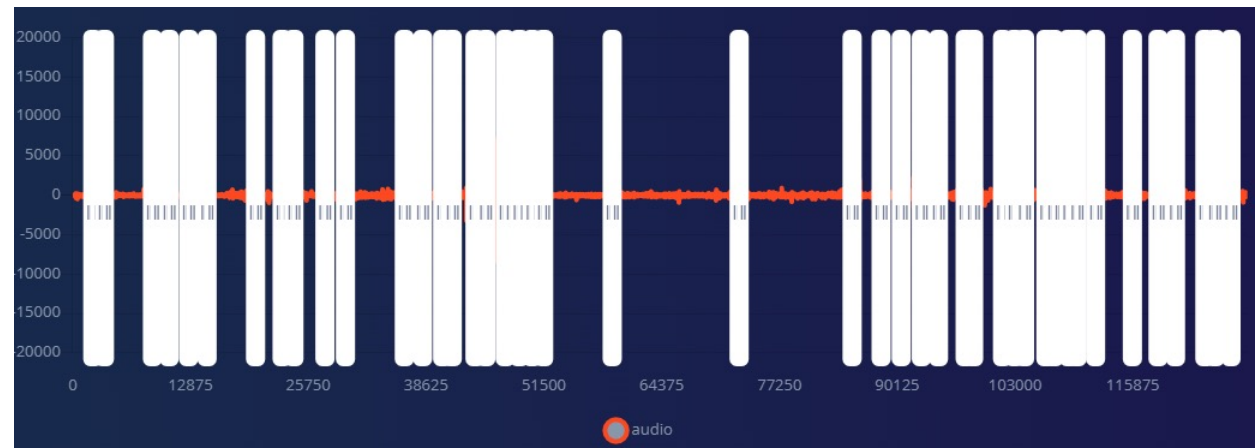
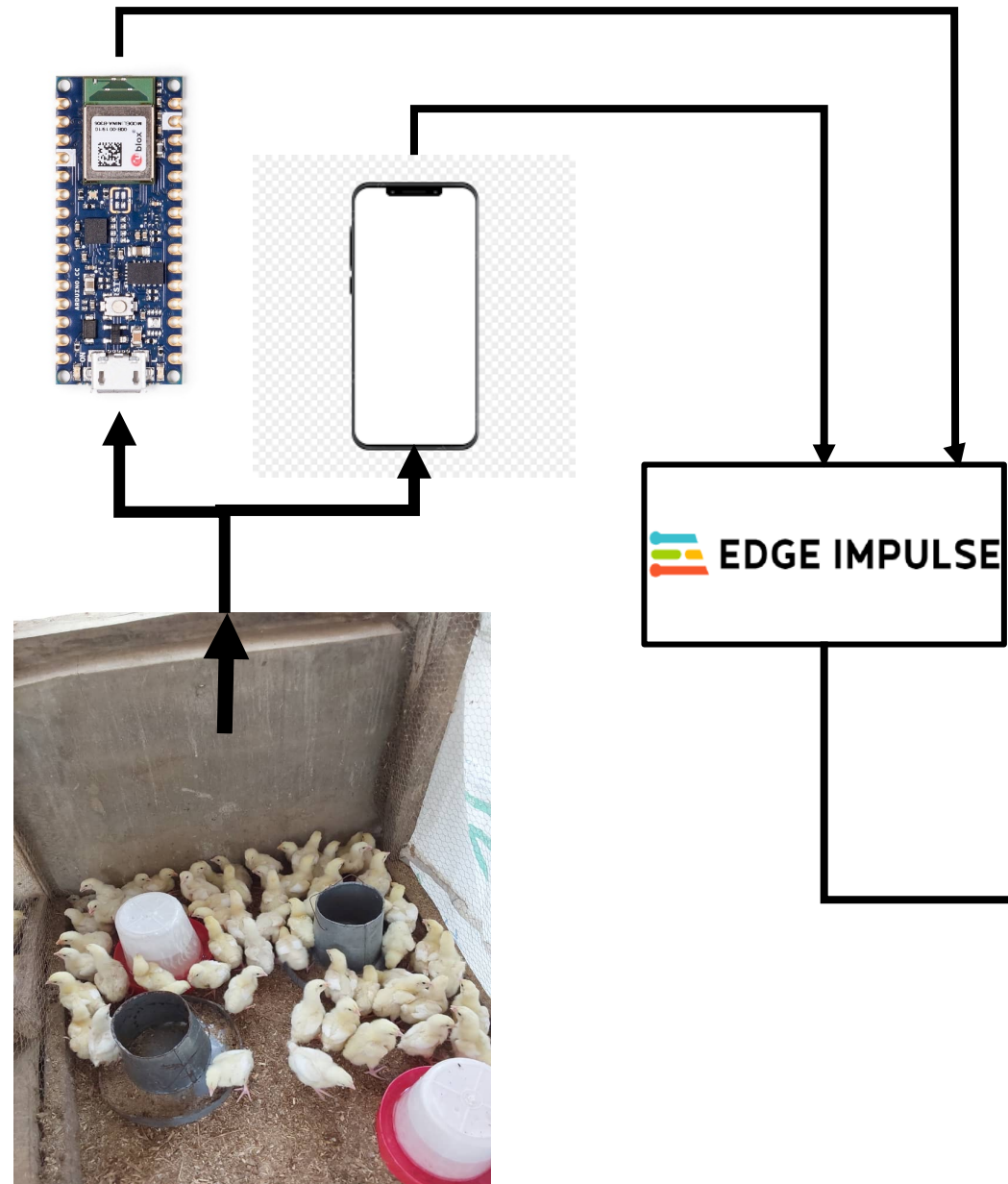
**2 WEEKS OLD**



**8 WEEKS OLD**



# DATA COLLECTION WITH SENSORS



- EDGE IMPULSE
- Dashboard
- Devices
- Data acquisition
- Impulse design
  - Create impulse
- EON Tuner
- Retrain model
- Live classification
- Model testing
- Performance calibration
- Versioning
- Deployment
- GETTING STARTED
  - Documentation
  - Forums

Segun Adebayo / Chicken Audio

Dataset | Data explorer | Data sources | CSV Wizard

DATA COLLECTED: 47s

TRAIN / TEST SPLIT: 0% / 100%

Collect data

[Connect a device](#) to start building your dataset.

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: Sick-bird-sound-1.s46



# DATASET DESCRIPTION

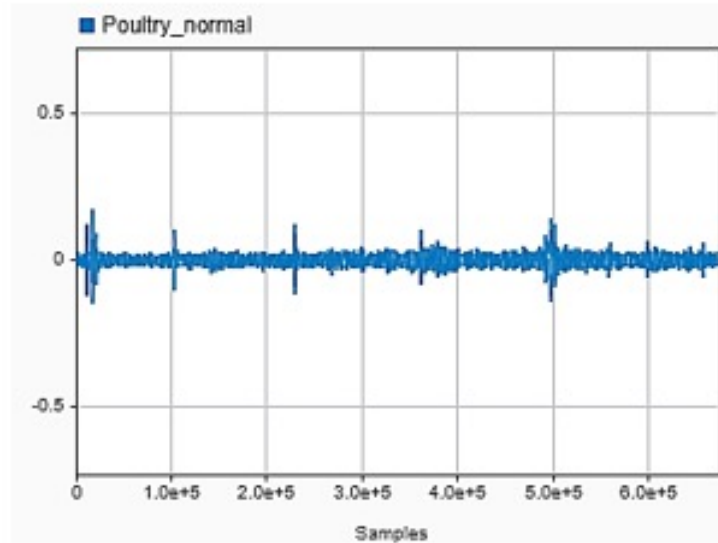
The dataset contains 346 audio signal files and grouped into three folders: healthy, noisy, and unhealthy.

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

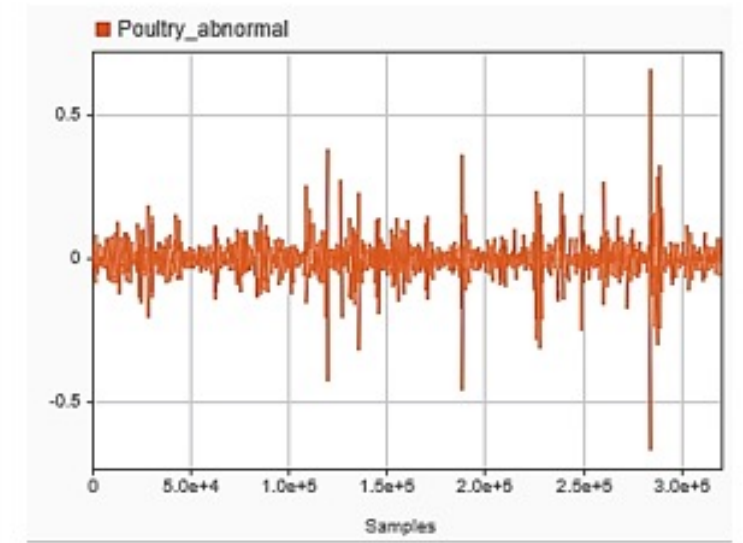
The audio signal file have a range of 5 seconds to 60 seconds length.

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

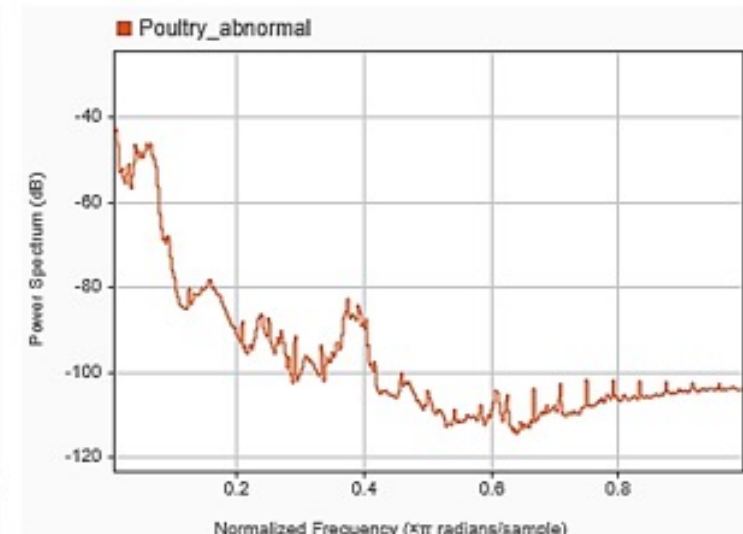
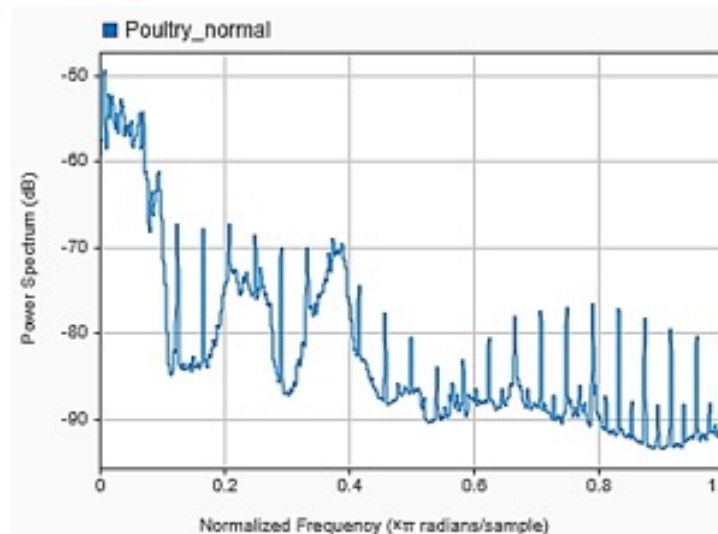
The Noise folder contain sound segments that include background noises (e.g., moving vehicle and human voices) and sound created by poultry bird activities (e.g., feeding, pecking one another).



(a)



(b)

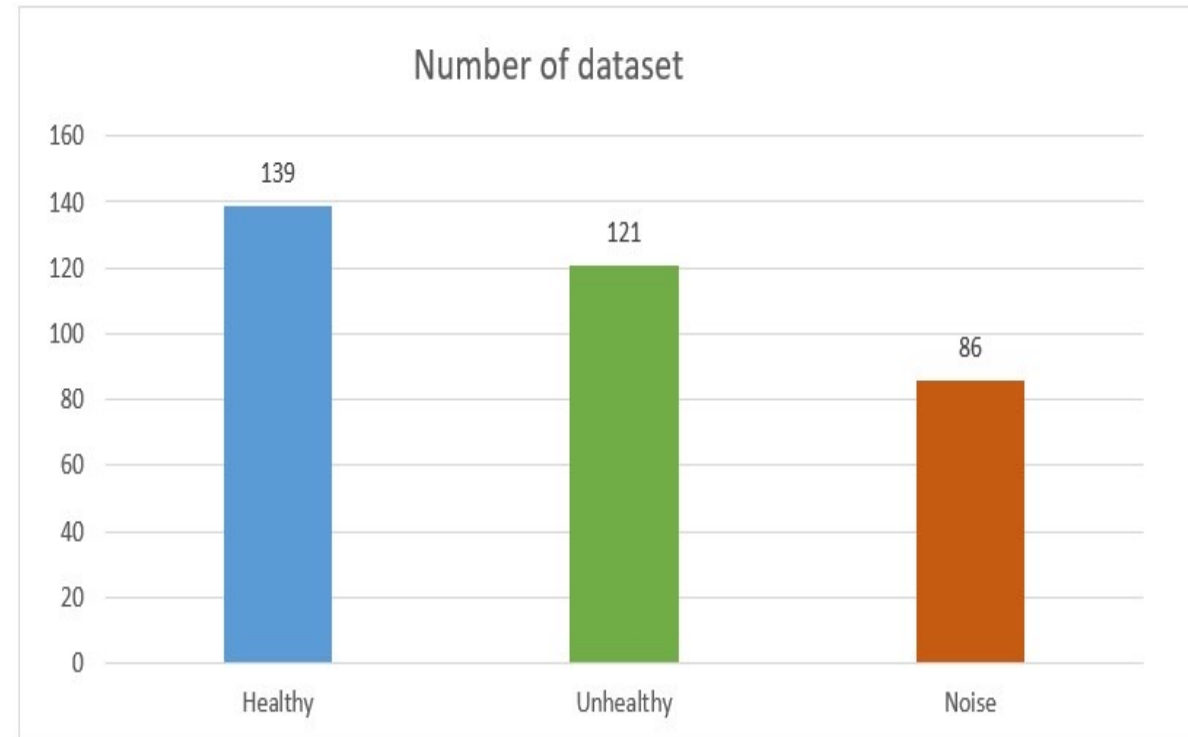


## Dataset Description

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.





# Poultry Vocalization Signal Dataset for Early Disease Detection

Published: 12 June 2023 | Version 1 | DOI: 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 65days. 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

## 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

[Copy to clipboard](#)



## Poultry Vocalization Signal Dataset for Early Disease Detection

Publication Year: 2023

715  
Usage

Home

Overview

Highlights



Metrics Details	
<b>USAGE</b>	<b>715</b>
Downloads	693
<a href="#">Mendeley Data</a>	693
Views	22
<a href="#">Mendeley Data</a>	22

### Dataset Description

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### Bibliographic Details

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**AUTHOR(S):**  
Aworinde, Halleluyah; Adebayo, Segun; Akinwunmi, Akinwale; Alabi, Olufemi; Ayandiji, Adebamiji; Oke, Olaide; Oyebamiji, Abel; Adeyemo, Adetunmbi; Aderonke, Echetama; Kizito

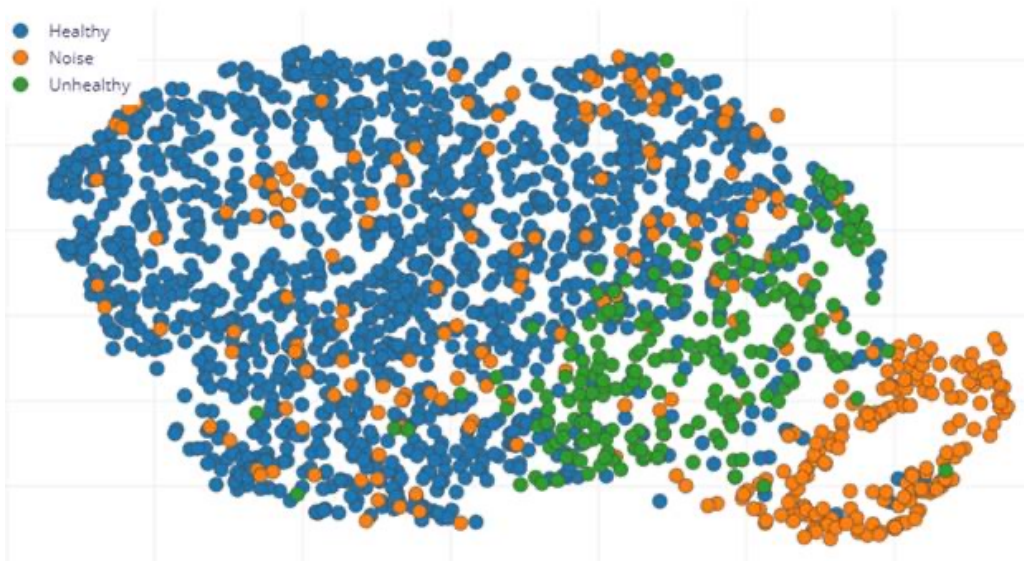
Show more



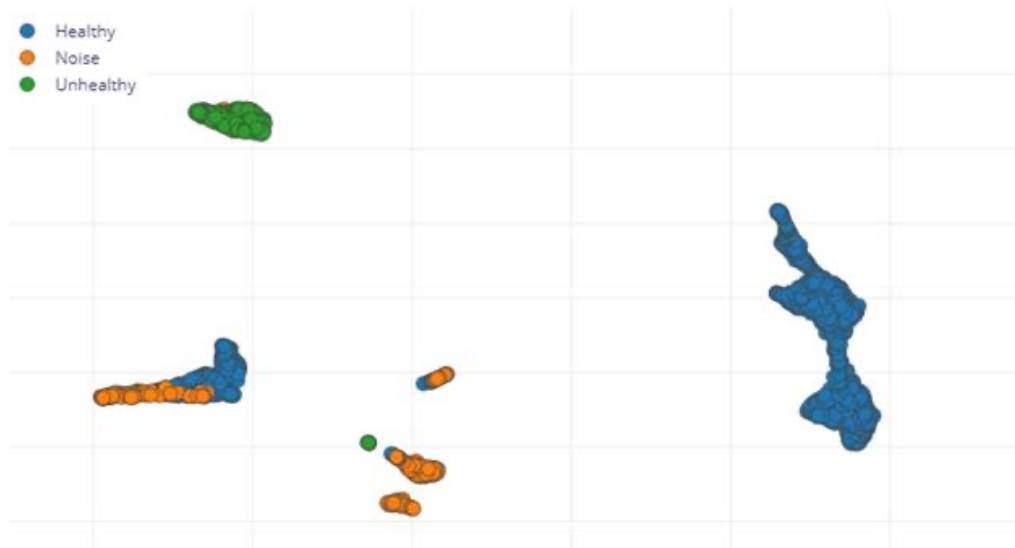


# INSIGHTS FROM DATASET FEATURES

MFCC Features



Spectrogram Features



MFE Features



# INSIGHTS FROM MODEL TRAINING

## Effect of Data Argumentation of Trained Model

<b>Add noise</b>	<b>None</b>	<b>Low</b>	<b>High</b>	<b>Low</b>	<b>High</b>	<b>High</b>	<b>High</b>
<b>Mask time bands</b>	None	None	None	Low	High	High	Low
<b>Mask frequency bands</b>	None	None	None	Low	High	None	None
<b>Accuracy</b>	96.4%	87.8%	80.2%	72.8%	39.0%	60.1%	57.8%
<b>F1-score</b>	96.3%	87.3%	79.6%	70%	28.3%	58%	55.3%
<b>Loss</b>	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

Model

Model version: ?

Quantized (int8) ▾

Last training performance (validation set)



ACCURACY  
99.2%



LOSS  
0.05

Confusion matrix (validation set)

	HEALTHY	NOISE	UNHEALTHY
HEALTHY	99.4%	0.4%	0.1%
NOISE	1.4%	98.3%	0.3%
UNHEALTHY	0.3%	0%	99.7%
F1 SCORE	0.99	0.99	1.00

Data explorer (full training set) ?



# Performance of Neural Network Training using MFE Features

Model

Model version: ?

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



**Dr. Halleluyah Aworinde**  
*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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THANK YOU...