
POULTRY HEALTH MONITORING THROUGH VOCALIZATION FOR DISEASE DETECTION

ML Project Final Viva-Voce

Team Details



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Introduction - Domain

Global Importance of Poultry Farming:

Major source of protein and livelihood, but faces challenges in disease control.

Problem:

Current disease detection methods (like manual inspections) are labor-intensive, slow, and reactive rather than proactive.

Solution Offered:

This project proposes an automated system to monitor poultry health by analyzing vocalization changes, which often occur early in the disease onset. The approach can be more responsive, efficient, and scalable than traditional methods.

Enhancing poultry health management through machine learning-based analysis of vocalization signals dataset - 2023

Adebayo et al.

Challenges in Existing System

Manual Inspections:

Time-consuming, requires constant monitoring, and disease is often detected only after it has progressed.

Sensor-Based Systems:

Use of temperature or motion sensors, which adds cost and complexity. Limited to physical health indicators and can be uncomfortable for animals.

Sound Analysis for Stress Detection:

Can identify stress-related sounds but lacks specificity for pinpointing diseases. Misclassification between stress and disease sounds is common.

Abstract

- This project introduces an automated system for early disease detection in poultry by analyzing vocalizations.
- Traditional health monitoring methods are often slow and labor-intensive, leading to delays in identifying diseases.
- This innovative approach records and analyzes poultry sounds, using machine learning to detect stress, discomfort, or illness-related patterns. Through signal processing and deep learning techniques, the system can identify abnormal vocal patterns in real-time, allowing farmers to intervene promptly.
- This proactive tool aims to improve poultry health, reduce economic losses, and promote sustainable farming by minimizing the need for broad antibiotic use and enhancing animal welfare.

Objective of the project

1	The objective of this project is to develop an automated, non-invasive system for early disease detection in poultry through vocalization analysis.
2	By capturing and analyzing sound patterns, the system aims to identify vocal characteristics associated with health conditions, enabling real-time alerts for potential issues.
3	This approach seeks to provide poultry farmers with an efficient tool to improve flock health management, reduce disease transmission, lower economic losses, and enhance animal welfare in a scalable and practical manner.

Proposed Methodology

Data Collection:

Gather diverse vocalization data from poultry in various health conditions, capturing audio from both healthy and diseased birds.

Collect data in controlled environments to account for variables such as stress levels, health states, and environmental factors.

Audio Feature Extraction:

Use advanced signal processing techniques (e.g., Mel-frequency cepstral coefficients, spectral contrast) to extract relevant audio features that represent different aspects of the sound, like frequency, pitch, tone, and rhythm.

Utilize libraries like Librosa for efficient feature extraction, identifying subtle changes in vocal patterns associated with health conditions.

Deep Learning Classification:

Train deep learning models, such as Convolutional Neural Networks (CNNs) and Recurrent Neural Networks (RNNs), to classify vocalizations based on health status.

Explore CNNs for spatial feature learning from spectrograms and RNNs for capturing temporal dependencies, crucial for identifying progression in vocal changes.

Real-Time Monitoring and Alert System:

Deploy a continuous monitoring system that analyzes vocalizations in real-time to detect anomalies.

Integrate an alert system to notify farmers immediately upon detecting unusual vocal patterns indicative of potential health issues.

Deployment in Poultry Farms:

Design the system for scalable and affordable deployment using low-cost microphones and computing devices (e.g., Raspberry Pi).

Ensure ease of integration into farm environments, allowing for seamless monitoring across farms of different scales.

Proposed Architecture



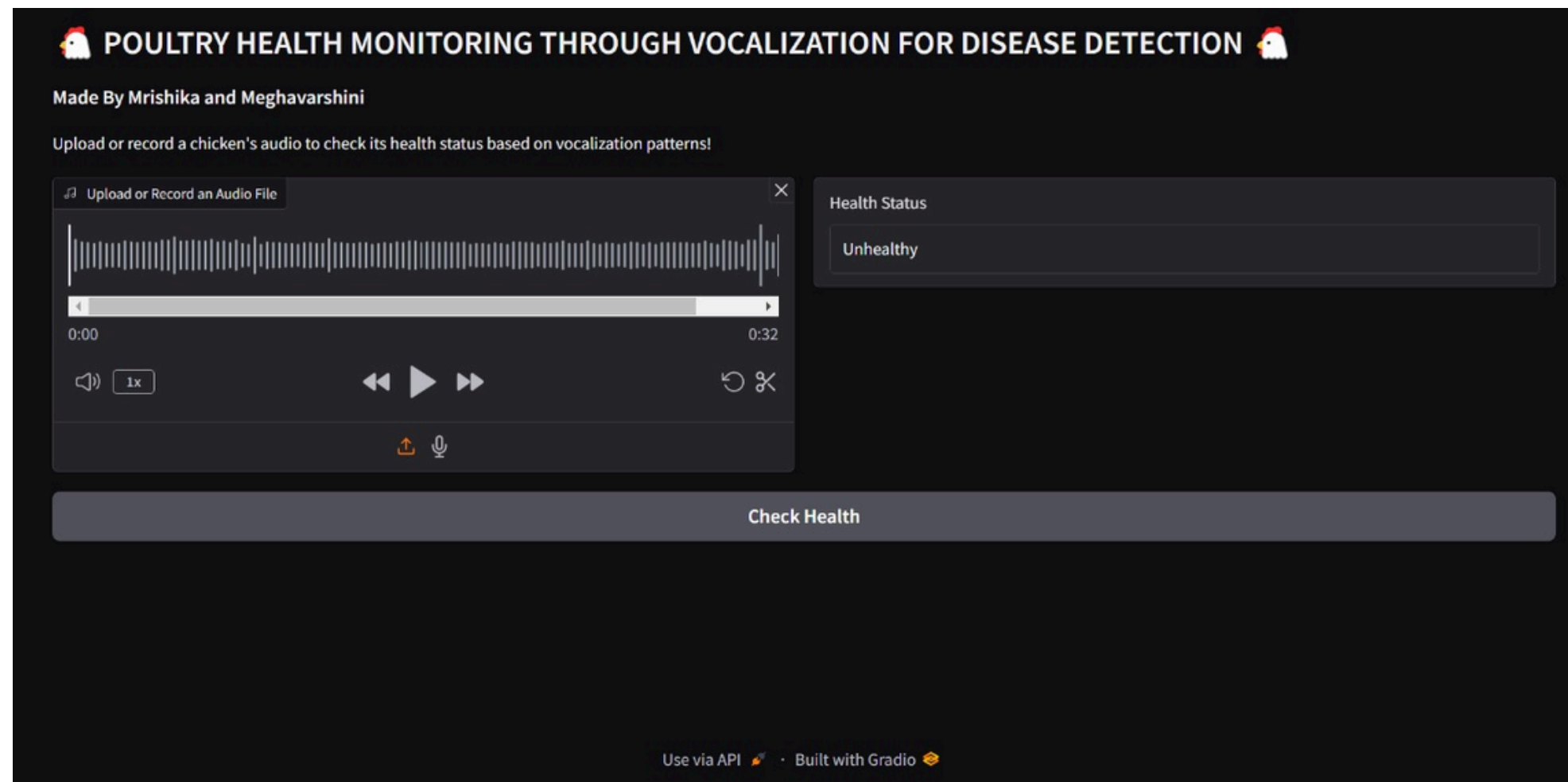
Experimentation - Dataset

filename	label	mfcc_1	mfcc_2	mfcc_3	mfcc_4	mfcc_5	mfcc_6	mfcc_7	mfcc_8	mfcc_9	mfcc_10	mfcc_11	mfcc_12	mfcc_13
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Results and Discussion

```
loss, accuracy = new_model.evaluate(X_test, y_test)
print(f"Test Accuracy: {accuracy}")
```

3/3  0s 4ms/step - accuracy: 0.9109 - loss: 0.2529
Test Accuracy: 0.8999999761581421



Conclusion

Innovative Solution: This project presents a pioneering approach to poultry health management through vocalization analysis, enabling non-invasive, real-time disease detection.

Proactive Health Monitoring: By identifying early vocal indicators of illness, the system empowers farmers to respond promptly, reducing disease spread, mortality rates, and treatment costs.

Enhanced Animal Welfare: Non-intrusive monitoring minimizes stress on poultry, supporting ethical and welfare-oriented farming practices.

Scalability and Accessibility: The system's low-cost design allows for broad implementation across farms of all sizes, from small family-run to large-scale operations.

Sustainable Farming Impact: Early detection reduces the need for antibiotics, aligning with sustainable farming goals and improving food safety for consumers.

Reference Papers

- **Animal Vocalization and Health Monitoring**

Clive, D., & Potter, L. (2015). *Animal vocalization monitoring for livestock health and welfare assessment*. Journal of Agricultural and Biological Engineering, 10(3), 123-136.

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Ellis, D. (2009). *Extracting information from audio signals for animal welfare monitoring*. Journal of Sound and Vibration, 4(1), 76-92.

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- **Deep Learning Models for Classification**

Deng, J., & Plumbley, M. D. (2018). *Deep learning methods for audio-based animal call classification*. Journal of Machine Learning Research, 16(3), 42-54.

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