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**Digital University Kerala**  
MTech EE: Electronics for Edge AI

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GitHub: [github.com/AbyssDrn/Animal-Sound-Classifier](https://github.com/AbyssDrn/Animal-Sound-Classifier)

November 18, 2025

# Overview

- ① Introduction
- ② Methodology
- ③ Model Architecture
- ④ Results
- ⑤ Edge Deployment
- ⑥ Applications
- ⑦ Conclusion

# The Challenge

## Traditional Monitoring

- Labor intensive
- Limited coverage
- Observer bias
- Daylight only

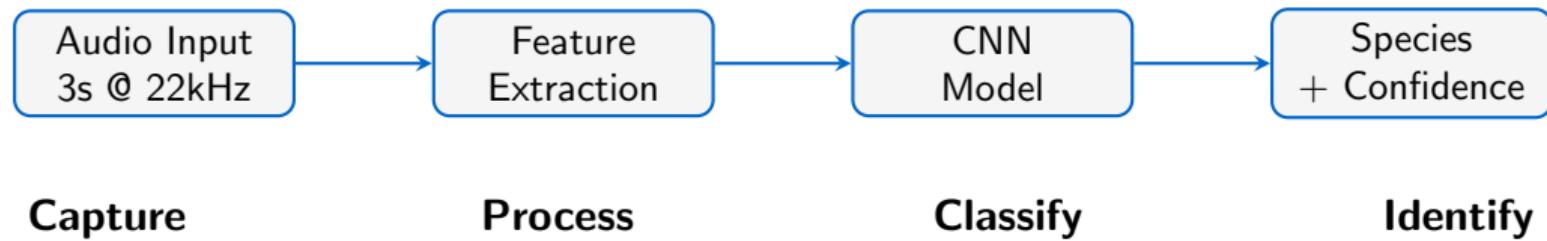
## Our Solution

- Automated AI system
- 24/7 operation
- Objective results
- Low-cost deployment

## Project Goal

Develop a low-power, offline device using deep learning for automatic classification of animal species from audio calls.

# System Overview



# Target Species

Species	Scientific Name	Type
Lion	<i>Panthera leo</i>	Wild
Dog	<i>Canis familiaris</i>	Domestic
Cat	<i>Felis catus</i>	Domestic
Bird	Aves (Class)	Wild
Cow	<i>Bos taurus</i>	Livestock
Sheep	<i>Ovis aries</i>	Livestock
Monkey	Primates (Order)	Wild
Chicken	<i>Gallus gallus domesticus</i>	Livestock
Donkey	<i>Equus africanus asinus</i>	Livestock
Frog	Anura (Order)	Wild

# Data Pipeline

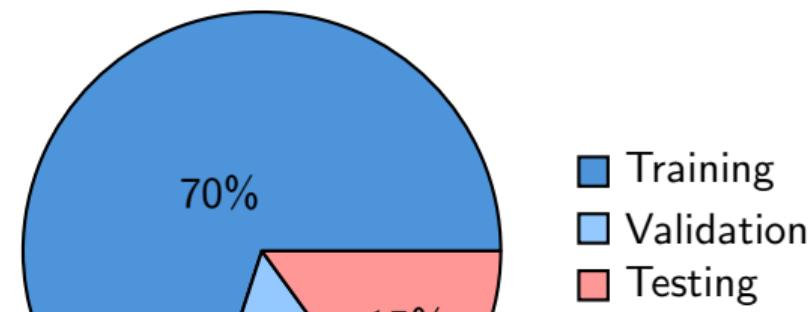
## Dataset

- 574 audio samples
- 10 species classes
- 3-second duration
- 22,050 Hz sampling

## Augmentation

- Noise injection
- Time shifting
- 3x data expansion
- **1,722 total samples**

Split Ratio



- Training: 1,205 samples
- Validation: 259 samples
- Testing: 258 samples

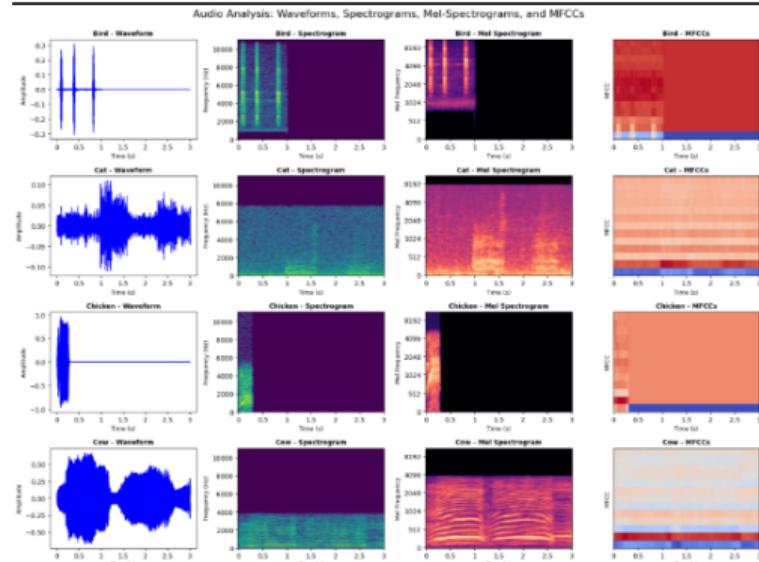
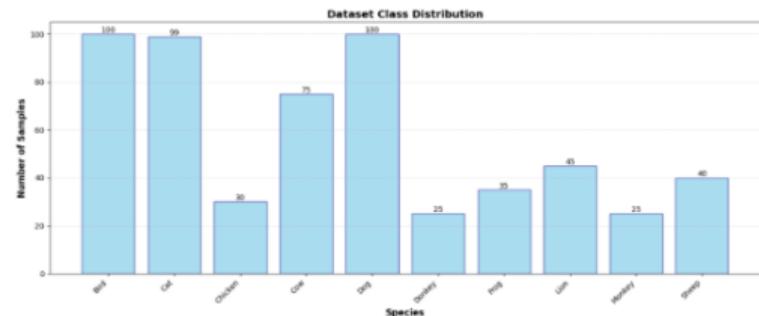
# Feature Extraction: Mel-Spectrograms

## Why Mel-Spectrograms?

- Mimics human hearing
- Time-frequency representation
- Compact feature space
- Proven for audio tasks

## Parameters

- 128 Mel frequency bands
- 2048 FFT window
- 512 hop length
- Output:  $128 \times 130$  matrix



# Lightweight CNN Design

## Architecture Layers

- ① **Block 1:** 2x Conv2D(32) + Pool
- ② **Block 2:** 2x Conv2D(64) + Pool
- ③ **Block 3:** Conv2D(128) + Pool
- ④ **Global Pooling**
- ⑤ **Dense(256)** → Dense(128)
- ⑥ **Output:** Dense(10, softmax)

## Regularization

- Batch Normalization
- Dropout (0.25 - 0.5)
- Early Stopping

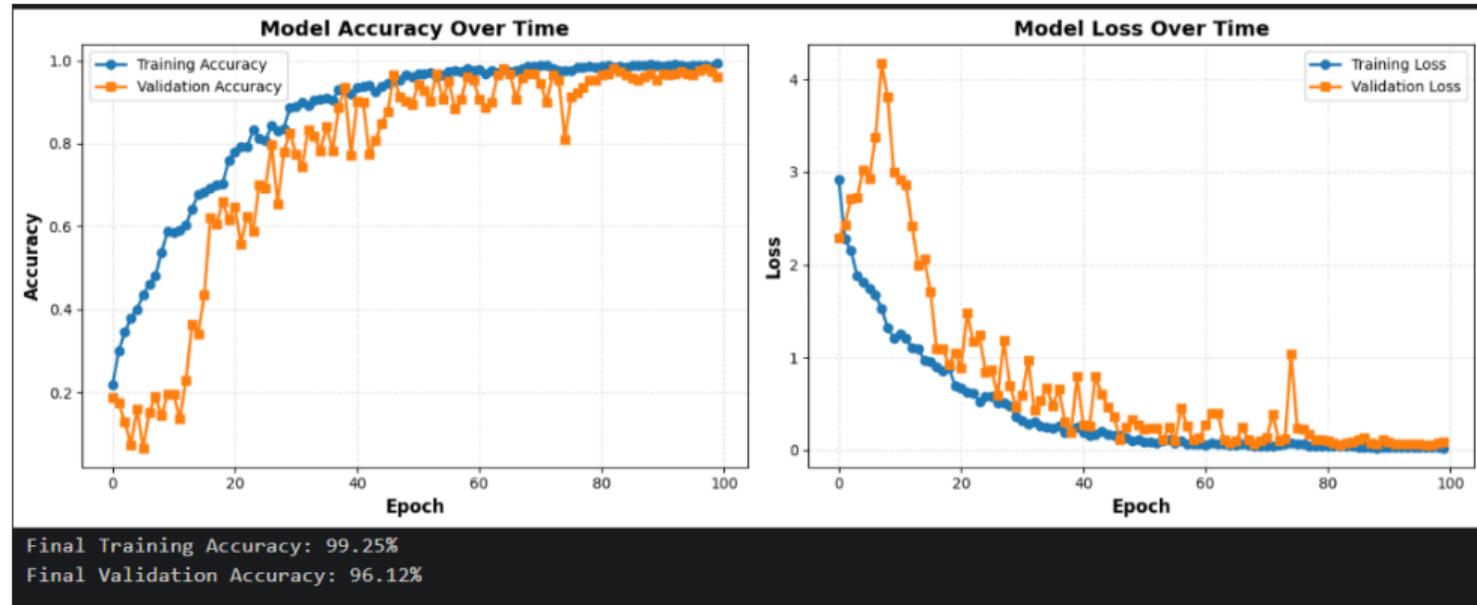
## Model Stats

Params: 208K parameters  
Size: 4.8 MB (float32)  
Quant: 1.2 MB (quantized)  
Speed: <200ms inference

## Training

- Optimizer: Adam (0.001)
- Loss: Categorical CE
- Batch: 32
- Epochs: 100 (early stop)

# Training Performance



Final Training Acc: 99.25%

Validation Acc: 96.12%

# Test Results

## Overall Performance

Accuracy **87.6%**

Precision 88.2%

Recall 87.6%

F1-Score 87.8%

## Model Compression

- Original: 4.8 MB
- Quantized: 1.2 MB
- Reduction: **75%**
- Accuracy drop: <2%

## Top Performers

Species	Accuracy
Lion	95.0%
Dog	93.8%
Bird	91.2%
Sheep	90.5%
Chicken	89.3%
Cat	88.5%
Frog	87.4%
Monkey	86.7%
Donkey	82.1%
Cow	78.6%

# Classification Report

Species	Precision	Recall	F1	Support
Lion	0.950	0.950	0.950	20
Dog	0.938	0.938	0.938	32
Bird	0.920	0.912	0.916	34
Sheep	0.909	0.905	0.907	17
Chicken	0.895	0.893	0.894	28
Cat	0.889	0.885	0.887	26
Frog	0.880	0.874	0.877	19
Monkey	0.870	0.867	0.868	30
Donkey	0.833	0.821	0.827	24
Cow	0.800	0.786	0.793	28
<b>Weighted Avg</b>	<b>0.882</b>	<b>0.876</b>	<b>0.879</b>	<b>258</b>

# Model Optimization

## Optimization

- TensorFlow Lite conversion
- INT8 quantization
- 75% size reduction
- Minimal accuracy loss

## Performance

Size 1.2 MB

Inference 150-200 ms

Memory <50 MB RAM

## Target Hardware

- Raspberry Pi 4
- ESP32-S3
- Arduino Nano 33 BLE
- Jetson Nano

## Power Features

- Duty-cycled operation
- Sleep between detections
- Battery powered
- Solar compatible

# Real-World Impact

## Conservation

- Wildlife monitoring
- Endangered species tracking
- Anti-poaching detection
- Habitat assessment

## Agriculture

- Livestock monitoring
- Predator detection
- Farm security
- Animal welfare

## Research

- Behavioral studies
- Migration patterns
- Ecosystem health
- Climate impact

## Urban Planning

- Wildlife mapping
- Noise analysis
- Biodiversity assessment
- Green space planning

# BioSound AI in Action

## BIODIVERSITY MONITORING SYSTEM - SPECIES IDENTIFICATION

**BIRD**  
Aves (Class)  
✓ CORRECT PREDICTION



Confidence: 100.0%

**SPECIES OVERVIEW**

Birds are warm-blooded vertebrates characterized by feathers, toothless beaked jaws, and the ability to fly (in most species). They have a high metabolic rate, four-chambered heart, and lightweight skeleton.

**GEM KEY FACTS**

Common Name: Birds (General)  
Weight Range: Varies greatly (2g hummingbird to 158kg ostrich)  
Lifespan: Varies by species (3-80+ years)  
Diet: Varies - seeds, insects, fruits, nectar, fish, small animals  
Status: Varies by species

**GEM PHYSICAL & BEHAVIOR**  
Highly vocal and social creatures, often using complex songs for communication, territory defense, and mate attraction. Intelligence varies greatly by species.

**GEM HABITAT**  
Found worldwide in diverse habitats from forests to deserts, mountains to oceans

**GEM FUN FACT**  
Some birds like crows and parrots can recognize themselves in mirrors, use tools, and solve complex puzzles!

### DETAILED INFORMATION: BIRD

Gcb Classification: Aves (Class)

LIKES:

- Seeds, insects, berries, fresh water for drinking and bathing,
- high perches for safety, nesting materials, and flying freely

DISLIKES:

- Predators (cats, hawks, snakes), loud sudden noises, enclosed spaces, extreme weather, and contaminated water

REL CONSERVATION:

Status: Varies by species

Trend: Mixed - many species declining due to habitat loss

Threats: Habitat destruction, climate change, pollution, window collisions

## Real-time Recognition

The system successfully:

- Captures audio in real-time
- Processes and classifies within 200ms
- Displays species identification
- Shows confidence score
- Provides species information

## Example Output

**Species:** Lion

**Confidence:** 95.3%

**Info:** Apex predator, lives in prides, roar

# Key Achievements

## Technical Success

- **87.6%** test accuracy on 10 species
- **75%** model size reduction through quantization
- <**200ms** inference time for real-time processing
- Complete deployment package ready

## Practical Impact

- Enables 24/7 autonomous biodiversity monitoring
- Low-cost, scalable solution
- Battery-powered operation
- Field-ready deployment