Transfer Learning-Based Classification of Poultry Diseases for Enhanced Health Management

1. Introduction

Project Title: Transfer Learning-Based Classification of Poultry Diseases for Enhanced Health Management

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Team size: 4

- **Team Leader: L Yuva shree**

Project Coordination, Milestone Tracking, Model Building, Final Validation

- **Team Member: N Charantejash**

– Data Collection, Cleaning, Preprocessing Pipeline

- **Team Member: Rodda Vamsi Krishna Reddy**

– Problem Definition, Problem Understanding

- **Team Member: Reddem Ganesh Reddy**

– ML Model Development & Flask Integration, UI/UX Design, Documentation

2. Project Overview

2.1 Purpose:

To develop an AI-driven system using transfer learning to identify common poultry diseases such as:

- Salmonella

- New Castle Disease

- Coccidiosis

- Healthy (no disease)

2.2 Features:

- Real-time image classification

- Symptom-based prediction

- Treatment suggestions

- Offline usability for rural farmers

- Educational resource for veterinary students

3. Architecture

3.1. Frontend

• Developed using **HTML5**, **CSS3**, and **Jinja2 templates** (via Flask)

• HTML form (index.html) allows users to **upload poultry images for prediction**

• UI is styled using embedded CSS and media served from the /static directory

• Clean and intuitive design for farmers and end-users

• Displays model prediction results (disease name) after image submission

3.2. Backend

• Backend logic implemented using **Python and Flask**

• **Model Training Pipeline:**

- Training, validation, and model export performed in **Google Colab**
- Data preprocessing, augmentation, and model training scripts run in Colab
- Final trained model exported as a .h5 file (e.g., healthy vs rotten.h5)

• **Model:**

A deep learning model using **Transfer Learning** (VGG16/VGG19/ResNet50)

Model artifacts include:

- poultry_model.h5– Trained model file
- Preprocessing logic embedded in the Flask app for real-time image handling

3.3. Data Preprocessing

• **Image resizing** (typically to 224x224 for VGG/ResNet compatibility)

• **Normalization** – Scaling pixel values to the range [0, 1]

• **Augmentation** – Flipping, rotating, and zooming during training to enhance performance

• Preprocessing logic handled during training and reused during inference in Flask backend

3.4. Database

• **No persistent database** is used

• All predictions are handled **in-memory**

• Model receives image input directly, processes it, and returns output without storage

• Optionally, uploaded images and results can be logged locally or stored in cloud storage if needed

4. Setup Instructions

4.1. Prerequisites:

a. Python 3.10+ or anaconda

b. Flask

c. TensorFlow or PyTorch

d. Google Colob

4.2. Installation

Git clone https://github.com/TrishaKanderi/Transfer-Learning-Based-Classification-of-Poultry-Diseases-forEnhanced-Health-Management cd poultry_disease_detection/Flask
pip install -r requirements.txt
python app.py

5. Folder Structure

5.1. Client (Flask Frontend)

The frontend is handled via **HTML templates** and **static assets**:

```
├── templates/
│   └── index.html          # Main HTML page for image upload and prediction
│
├── static/
│   └── images/
│       └── poultry-bg.jpg # Background image
│       └── healthy.png     # Image used for healthy prediction
│       └── infected.png    # Image used for infected result
```

- index.html contains a **form to upload poultry images** and a **predict** button.
- **images** used for UI styling are stored under the static/ folder.

5.2. Server (Flask Backend)

The backend is developed using **Python and Flask**:

```
├── app.py                # Main Flask application file (routing & prediction)
├── poultry_model.h5      # Trained CNN model (VGG16/VGG19/ResNet50)
├── preprocess.py         # (Optional) Image preprocessing helper functions
├── train_model.ipynb     # Google Colab notebook for model training
```

├── tensorflow.txt # Python package dependencies

- app.py handles:
  - Routing (/ and /predict)
  - Loading the .h5 model
  - Image preprocessing and prediction logic
- poultry_model.h5 is the **trained deep learning model** saved from Colab.
- train_model.ipynb includes:
  - Dataset loading
  - Data preprocessing and augmentation
  - Model training and evaluation

6. Running the Application

- Start the app:

  cd Flask

  python app.py

- Open in browser at: http://127.0.0.1:5000/

7. API Documentation

This is a form-based web application. The /predict route accepts POST requests from the image upload form and returns an HTML page with the prediction result.

POST /

Route: / (root URL)
Method: POST

Input:

- User uploads a poultry image (e.g., image of a sick chicken) via the HTML form
- File input field: name="file"

Processing:

1. The uploaded image is:
   - Loaded and resized (typically to 224x224)
   - Normalized (pixel values scaled to [0,1])
   - Converted into an array format compatible with the model
2. The processed image is passed to the trained model (poultry_model.h5) for classification.
3. The model predicts the type of poultry disease (e.g., Newcastle Disease, Avian Influenza, etc.)

Output:

- Rendered index.html page displaying:
  - Prediction result: e.g., "Predicted Disease: Newcastle Disease"
  - Visual feedback: Image preview and result text

Error Handling:

- Displays a friendly error message if:
  - No image is uploaded
  - An unsupported file type is uploaded
  - File is too large or unreadable
- Flask app includes basic validations to ensure smooth user experience

8. Authentication

Not applicable – This application does not implement user login or authentication as it's a prototype for medical prediction.

9. User Interface

- Simple form to upload poultry bird images for disease prediction
- Clean UI with a semi-transparent poultry-themed background
- Responsive layout with easy-to-use buttons
- Image preview with the result

10. Testing

- Model accuracy validated using evaluation metrics (accuracy, precision, recall) in Google Colab
- Manual testing of image upload form and prediction flow via Flask UI
- Model tested on a holdout test set (80/20 train-test split) to ensure generalization
- Verified predictions across different poultry disease classes for consistency

11. Screenshots or Demo

Demo video link:
https://drive.google.com/file/d/1TWK9cS-3sLZSRlMsyHsPNICjiTRjSVwn/view?usp=drivesdk

12 Known Issues

Limited dataset impacts prediction accuracy

No multi-language support (planned)

Poor image quality reduces model performance

13 Future Enhancements

Add more poultry diseases (e.g., Avian Influenza)

Integrate chatbot support for farmers

IoT sensor-based environment tracking

Admin dashboard for farm-wide monitoring