

ENCHANTED WINGS: MARVELS OF BUTTERFLY SPECIES

1. INTRODUCTION

Project Title: Enchanted Wings: Marvels of Butterfly Species

Team Members:

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 - Polavarapu Ramani – Frontend Developer
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2. PROJECT OVERVIEW

Purpose:

This project aims to develop a robust image classification system for identifying butterfly species using deep learning. It uses transfer learning with VGG16 and a Flask-based web app to deliver real-time predictions from user-uploaded images.

Features:

- Upload butterfly image and classify its species
 - Display prediction confidence
 - Web-based interface using Flask and Bootstrap
 - Responsive layout for desktop and mobile use
 - Visual performance metrics (accuracy graph)
 - Integrated model training in Google Colab
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3. ARCHITECTURE

Frontend:

- Built with HTML5, Bootstrap 5, and JavaScript
- Responsive UI for displaying prediction results and forms for uploading images
- Templates for homepage (carousel), input form, and result output

Backend:

- Flask (Python) backend
- TensorFlow/Keras used for loading and running VGG16-based classifier
- Pretrained model stored as .h5 file and loaded into memory on app start

Database:

- No persistent storage used. All images are processed temporarily in-memory during prediction
 - Option to add MongoDB in future for storing user submissions and prediction logs
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4. SETUP INSTRUCTIONS

Prerequisites:

- Python 3.8+
- Flask
- TensorFlow
- Keras
- Google Colab (for model training)

Installation:

1. Clone the project repo
 2. Place your vgg16_butterfly_final.h5 model file in the project root
 3. Install dependencies using `pip install -r requirements.txt`
 4. Run `python app.py` to start the Flask server
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5. FOLDER STRUCTURE

PROJECT_FOLDER/

```
|
|— main.py
|— vgg16_butterfly_model.keras
|— butterfly_classification.ipynb
|
|— templates/
|   |— index.html
|   |— result.html
|— requirements.txt
```

6. RUNNING THE APPLICATION

Frontend: Automatically rendered by Flask from `/templates/index.html`

Backend: Start with:

```
python app.py
```

Access the application at `http://localhost:5000`

7. API DOCUMENTATION

Endpoint: /predict

Method: POST

Request: Multipart Form Data (image upload)

Response: JSON with prediction result

Example:

```
{
  "prediction": "Monarch Butterfly",
  "confidence": 0.94
}
```

8. AUTHENTICATION

This version does not require user authentication. Future enhancements may include:

- User registration/login
- Token-based authentication
- Prediction history for logged-in users

9. USER INTERFACE

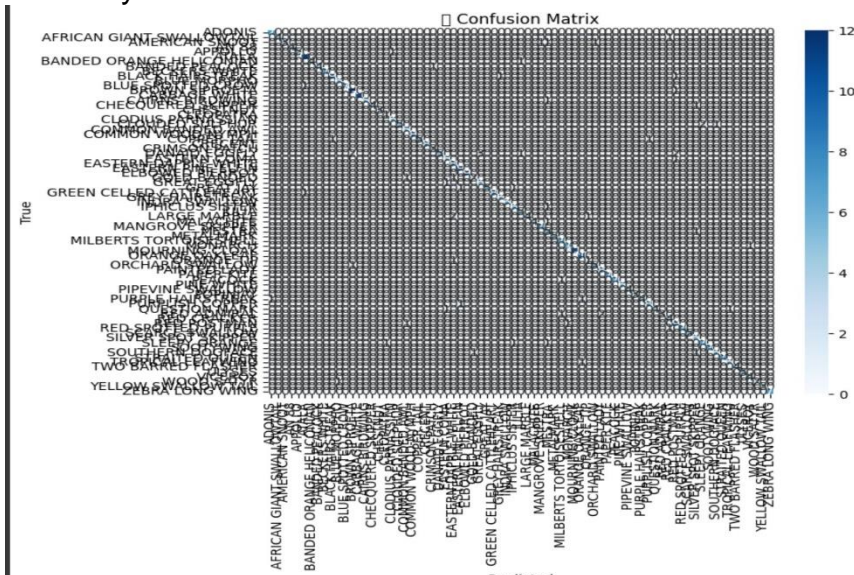
- Homepage: Carousel with “Get Started” button leading to the input form
- Input Page: Users upload a butterfly image and submit
- Output Page: Displays the predicted species and confidence

10. TESTING

- Functional Testing for all endpoints
- UI Testing for responsiveness and layout
- Model evaluation using metrics like Accuracy, Precision, Recall
- Accuracy on training set: **96.4%**

accuracy			0.88	500
macro avg	0.90	0.88	0.87	500
weighted avg	0.90	0.88	0.88	500

- Accuracy on validation set: **88.2%**



16/16 ————— 11s 693ms/step

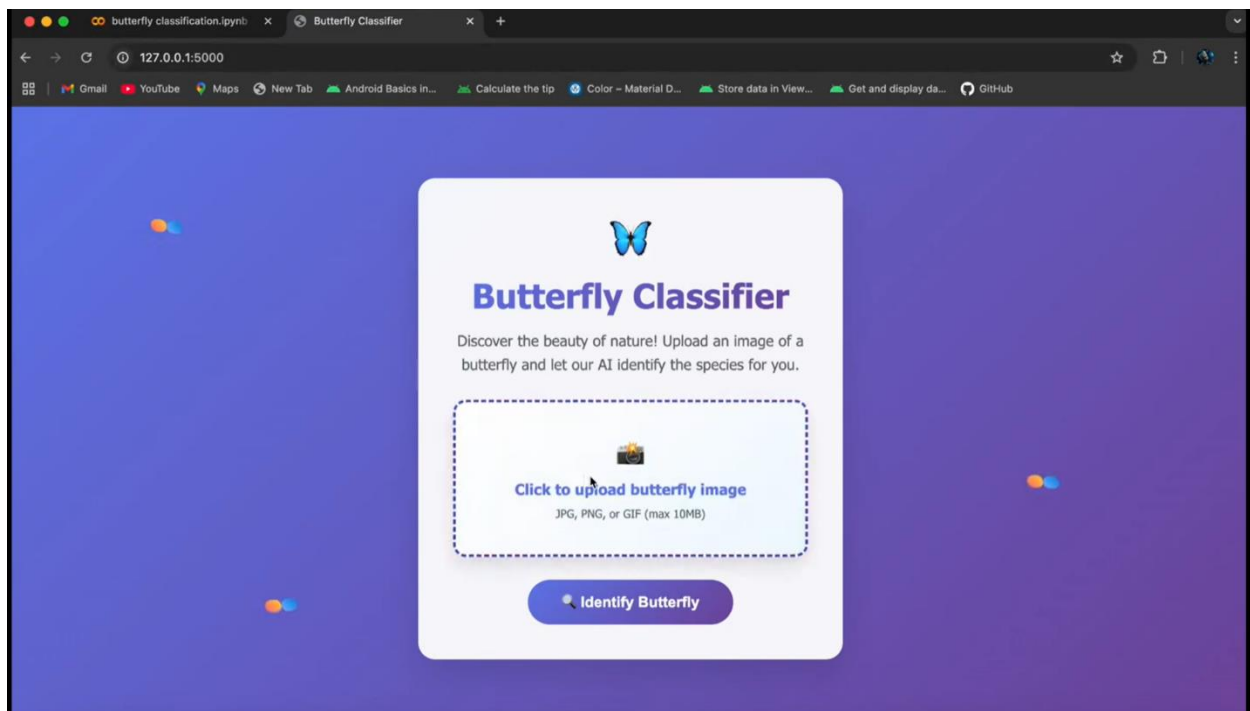
✓ Accuracy: 87.80%

📊 Classification Report:

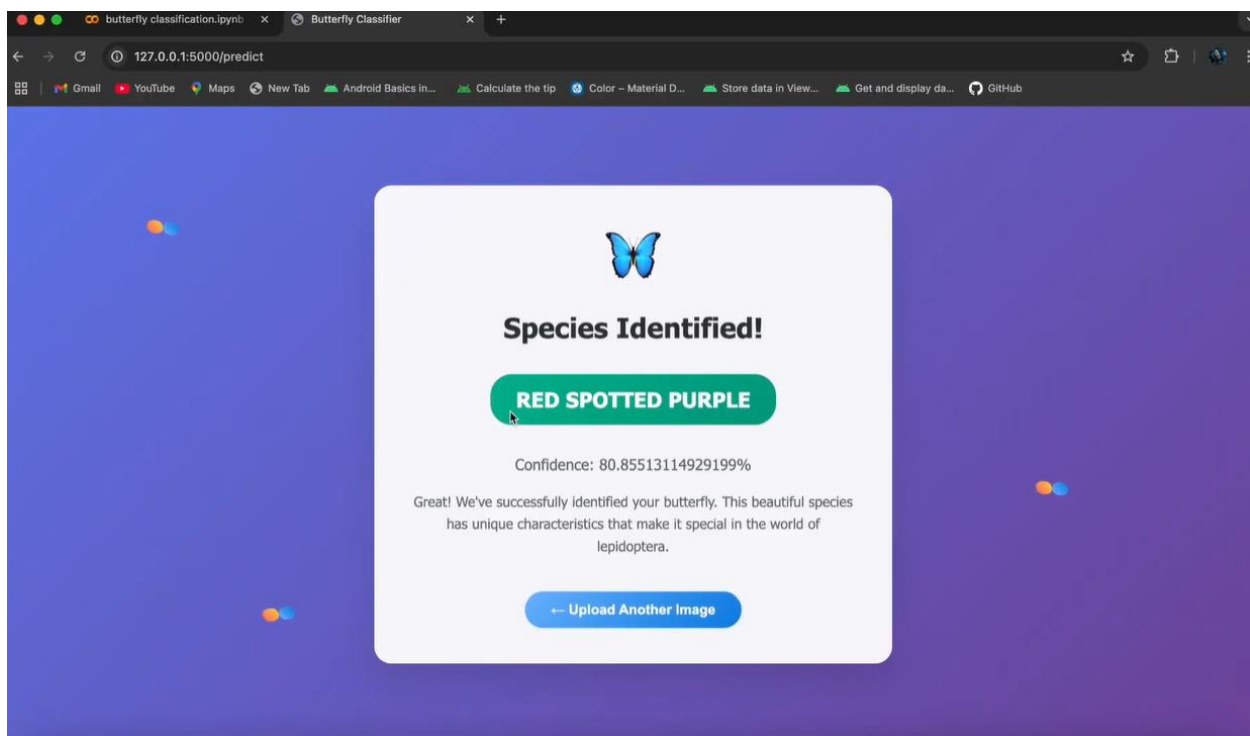
	precision	recall	f1-score	support
ADONIS	0.88	1.00	0.93	7
AFRICAN GIANT SWALLOWTAIL	1.00	0.80	0.89	5
AMERICAN SNOOT	1.00	0.50	0.67	6
AN 88	1.00	1.00	1.00	4
APPOLLO	1.00	0.83	0.91	6
ATALA	0.85	1.00	0.92	11
BANDED ORANGE HELICONIAN	1.00	0.80	0.89	5
BANDED PEACOCK	1.00	0.80	0.89	5
BECKERS WHITE	1.00	1.00	1.00	7
BLACK HAIRSTREAK	0.90	0.82	0.86	11
BLUE MORPHO	0.86	1.00	0.92	6
BLUE SPOTTED CROW	1.00	0.88	0.93	8
BROWN SIPROETA	0.80	0.92	0.86	13
CABBAGE WHITE	1.00	1.00	1.00	11
CAIRNS BIRDWING	1.00	0.86	0.92	7
CHECQUERED SKIPPER	1.00	0.80	0.89	5
CHESTNUT	1.00	1.00	1.00	3

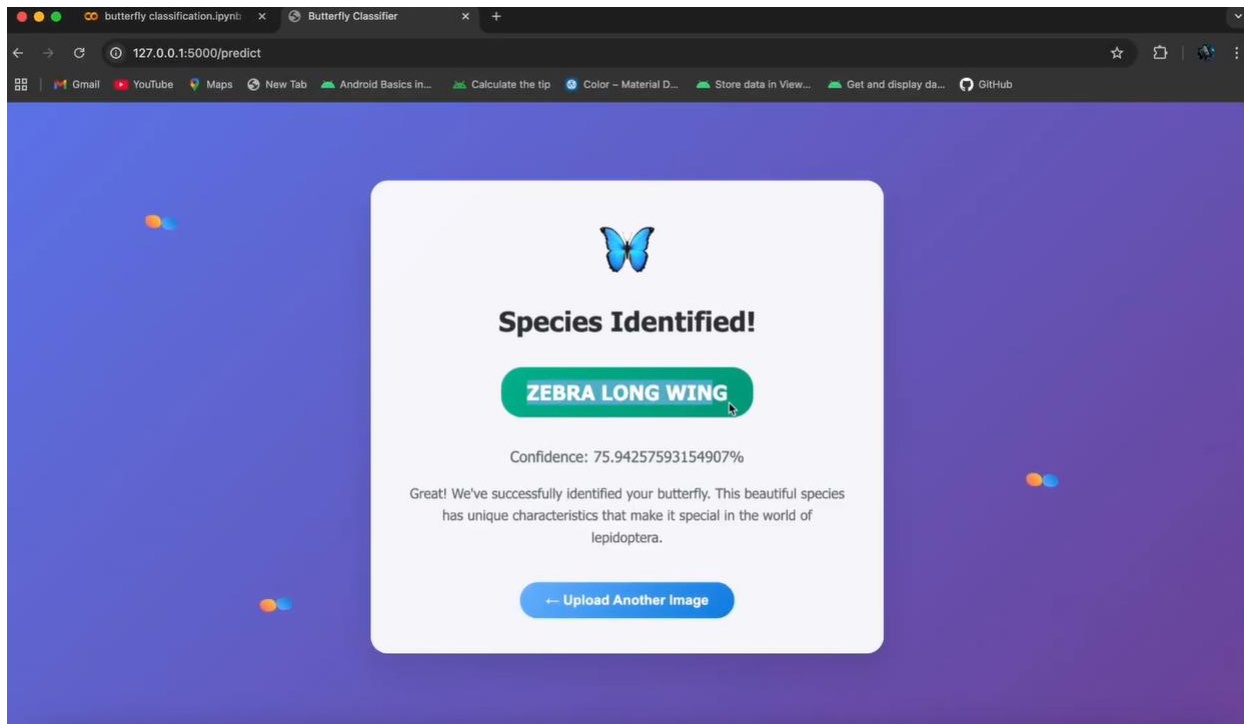
11. SCREENSHOTS OR DEMO

Homepage: Carousel layout with navigation bar



Output Page: Prediction result with formatted species name and confidence bar





Demo Video link-- https://drive.google.com/file/d/17mWkBUcEhKNNNr-gzKDizvduP7wpkjh7/view?usp=drive_link

12. KNOWN ISSUES

- Limited real-time camera input support
- Misclassification for rare species
- Model performance depends heavily on image quality
- Prediction is slightly slower on low-end hardware

13. FUTURE ENHANCEMENTS

- Add mobile camera integration
 - Migrate backend to FastAPI for better performance
 - Use cloud inference with Google Cloud Functions or AWS Lambda
 - Add voice-based interaction (speech to species)
 - Enhance dataset with community submissions
 - Add visualization for prediction confidence overlaid on the image
 - Include more species and retrain the model for wider accuracy
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