



# Intelligence Academy

The Center for Research and Development



## Butterfly Image Classification Capstone Project

Automated Butterfly Species Identification Using Deep Learning



### Project Assessment Criteria

Section	Marks (%)	Description
Introduction & Background	10%	Problem statement, objectives, and significance.
Literature Review	10%	Review of existing solutions and deep learning techniques.
Dataset & Preprocessing	15%	Data understanding, augmentation, and preprocessing steps.
Methodology (Model Design)	20%	CNN architectures, training process, and hyperparameter tuning.
Evaluation & Results	15%	Model performance analysis using metrics, visualizations.
Deployment & Web Application	10%	API development and UI implementation for real-world use.
Discussion & Conclusion	10%	Error analysis, limitations, and future improvements.
Report Quality & Presentation	10%	Clarity, formatting, documentation, and references.



### 1. Introduction

## 1.1 Background

Butterflies are essential to ecosystems, playing a role in pollination and indicating environmental health. However, manual classification is **challenging** due to:

- **Similar wing patterns** across species.
- **High intra-class variation** due to environmental conditions.
- **Manual expertise requirement** for accurate identification.

## 1.2 Problem Statement

This project aims to **automate butterfly species identification** using deep learning, particularly Convolutional Neural Networks (CNNs). The goal is to classify **75 different butterfly species** from images.

## 1.3 Objectives

- Train a deep learning model to **classify butterfly species**.
- Improve classification accuracy using **transfer learning and augmentation**.
- **Deploy the model** as a web-based application for real-time predictions.

## 1.4 Significance

- **Biodiversity Conservation:** Helps researchers identify rare species.
- **Citizen Science:** Enables public participation in species tracking.
- **Automated Research Tool:** Supports ecological and entomological studies.



## 2. Literature Review (10%)

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### 2.1 Traditional Approaches

- **Manual identification** by taxonomists.
- **Feature-based ML models** (SVM, k-NN) but require handcrafted features.

### 2.2 Deep Learning in Image Classification

- **CNNs (Convolutional Neural Networks)** dominate image recognition tasks.
- Pretrained architectures like **ResNet, VGG, and EfficientNet** improve accuracy.
- **Data Augmentation & Fine-Tuning** enhance performance in small datasets.

## 2.3 Existing Research

- Insect identification models have reached 85-90% accuracy with CNNs.
- Most lack class diversity or are not real-world deployable.



## 3. Dataset & Preprocessing (15%)

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### 3.1 Dataset Overview

- Source: Dataset consists of 1000+ labeled images of 75 butterfly species.
- Train/Test Split:
  - Training images: **Training\_set.csv**
  - Testing images: **Testing\_set.csv** (unlabeled images for prediction).

### 3.2 Preprocessing Steps

- Image Resizing: All images resized to 224x224 pixels for consistency.
- Normalization: Pixel values scaled between [0,1].
- Data Augmentation:
  - Rotation ( $\pm 20^\circ$ ), Flipping, Brightness Adjustments.
  - Synthetic images generated for underrepresented classes.

### 3.3 Exploratory Data Analysis

- Class Distribution Visualization (Identify data imbalance).
- Sample Image Grid (Show representative species).
- Outlier Detection (Remove corrupted images).



## 4. Methodology (Model Design) (20%)

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### 4.1 Model Selection

We test various CNN architectures:

1. Baseline CNN: 3 Conv layers, Pooling, Dense layers.
2. Pretrained Models:
  - VGG16
  - ResNet-50
  - EfficientNetB0

### 3. Fine-tuned Model:

- Best architecture is **fine-tuned** using **Transfer Learning**.

## 4.2 Model Architecture

- **Input Layer:** Accepts 224x224 images.
- **Convolutional Layers:** Extract hierarchical features.
- **Batch Normalization:** Stabilizes training.
- **Dense Layer:** Final classification into **75 classes**.
- **Activation Function:** Softmax for multi-class classification.

## 4.3 Training Setup

- **Loss Function:** Cross-Entropy Loss.
- **Optimizer:** Adam / SGD with momentum.
- **Learning Rate Schedule:** ReduceLROnPlateau.
- **Hyperparameter Tuning:** Grid search for best dropout rate, batch size.

## 4.4 Model Training

- Trained for 50 epochs with early stopping.
- GPU-accelerated training on **Google Colab / Kaggle Notebooks**.



## 5. Evaluation & Results (15%)

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### 5.1 Performance Metrics

- **Accuracy:** Measures overall correctness.
- **Precision & Recall:** Balance between false positives & false negatives.
- **F1-Score:** Best metric for **imbalanced classes**.

### 5.2 Confusion Matrix

- Visualize misclassifications to identify problematic classes.

### 5.3 Model Comparison

Model	Accuracy	F1-Score
Baseline CNN	72%	0.70

Model	Accuracy	F1-Score
ResNet-50	85%	0.84
EfficientNetB0	91%	0.90

## 5.4 Error Analysis

- Investigate misclassified images.
- Study challenging species with overlapping patterns.

## 6. Deployment & Web Application (10%)

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### 6.1 API Development

- Framework: FastAPI / Flask.
- Model hosted on Hugging Face Spaces / Google Cloud.

### 6.2 Web Interface

- Built with Streamlit for an interactive UI.
- Features:
  - Upload an image.
  - Get real-time butterfly species prediction.
  - Display top-5 similar species.

### 6.3 Docker & Cloud Hosting (Optional)

- Containerized using Docker.
- Deployment on AWS/GCP for accessibility.

## 7. Discussion & Conclusion (10%)

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### 7.1 Key Findings

- Achieved 91% accuracy using EfficientNet.
- Fine-tuning improved generalization.

### 7.2 Limitations

- Small dataset; **could benefit from synthetic data.**
- **No multi-angle images** (only frontal views).

## 7.3 Future Improvements

- **Use Self-Supervised Learning** for better feature extraction.
- **Mobile App Deployment** for field research.



## 8. References (10%)

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- Research papers on **CNN-based species classification.**
- Official documentation of **TensorFlow, PyTorch.**
- Biodiversity studies on butterfly species.



## Final Submission Checklist

- ✓ **Code Repository** (GitHub link).
- ✓ **Final Report** (PDF).
- ✓ **Web App Link** (Deployed Model).
- ✓ **Presentation Slides** (Capstone Defense).