

OBJECTIVE

Create a modular solution for classifying a bird dataset, using Docker and RabbitMQ for communication between modules.

Key Components:

- Producer: Generates raw image data.
- Processor: Transforms raw data into training-ready format.
- Uploader: Stores processed data in MongoDB.
- Receiver: Retrieves images for training from DB.
- Trainer: Trains the model with new data.
- Predictor: Uses the latest model for making predictions.
- Model Uploader: Manages model upload in chunks.
- App: Provides a web interface for interaction.
- RabbitMQ: Manages asynchronous messaging between components, ensuring reliable data flow.
- MongoDB: Uses GridFS for scalable storage of images and models.

RABBITMQ INTEGRATION

Function: Manages communication between components through various queues.

Reason: Enables asynchronous, reliable message passing, allowing components to operate independently. (And requirements ©)

Benefits

- Decoupling: Components can work independently and communicate via messages.
- Scalability: System can handle increased load by managing message queues effectively.

MONGODB USAGE

Benefits

- **Scalability**: Handles large datasets efficiently.
- **Flexibility**: Easily stores and retrieves large files.

GridFS

Function: Stores images and models using GridFS for handling large files.

Reason: Traditional BSON document storage limits are overcome by GridFS, which is designed for large data.

DEPLOYMENT IN DOCKER

Benefits

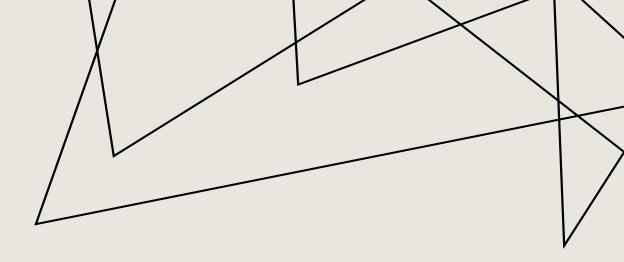
- **Flexibility**: Easily deployable in various environments.
- Efficiency: Can scale components as needed to handle increased workload.

Scalability

- Load Handling: Designed to handle large volumes of data and scale out for more compute resources.
- Modular Design: Each component can be scaled independently based on load.

CHALLENGES

- Data Volume: Handling and processing large volumes of image data efficiently.
 Especially in RabbitMQ
- **Model Integrity**: Ensuring the accuracy and completeness of model uploads and downloads.
- Birds and clouds: You will see 😂 (We still hate frontend)



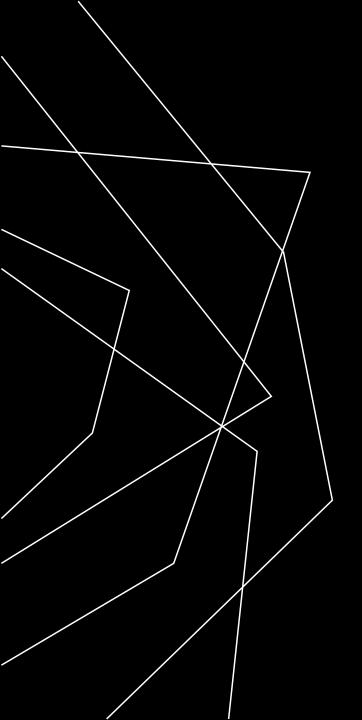
WHY INCEPTIONV3?

Practice makes perfect

- Architecture: InceptionV3 is a deep convolutional neural network known for its efficiency and accuracy.
- Pre-trained Weights: Provides a strong starting point for transfer learning, leveraging pre-trained weights from ImageNet.
- **Suitability**: Effective for image classification tasks, making it ideal for the bird species classification in this project.

Benefits

- **Efficiency**: Reduces training time and computational resources.
- Accuracy: Achieves high performance in image classification tasks.
- **Flexibility**: Adaptable to various image-related tasks through transfer learning.



WORKFLOW AND DEMO