

# About Project

## □ About the Project

### 🔍 Purpose

This project focuses on **automated flaw detection in Non-Destructive Testing (NDT)** using deep learning. It helps identify **flaw (defect)** and **no-flaw (normal)** regions in industrial inspection images (ultrasonic, X-ray, etc.) to improve quality, safety, and throughput.

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## □ Dataset

We use the **Koomas NDT\_ML\_Flaw** dataset. Data is extracted into `.npy` shards (`images.f16.npy` and `labels.u1.npy`) with a `manifest.json` that indexes parts and counts for efficient IO.

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## □ Model Overview

A CNN is trained (via `train_model.py`) to classify each image strip as:

- `1` → Flaw
- `0` → No Flaw

Preprocessing includes normalization and cropping a specific flaw band ( $\approx 1100$ – $3100$  px from a  $7168$ -px width), mirroring training/inference for consistency. Models and artifacts are saved in `models/` (`best.keras`, optional `threshold.txt`, `training_log.csv`, `metrics.txt`).

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## ⚡ Prediction Pipeline

1. Load the trained model (`best.keras` / `final.keras`).
  2. Preprocess (crop, resize, normalize) each strip.
  3. Output a **probability** for *flaw*.
  4. Apply a configurable **decision threshold** to classify **Flaw** vs **No Flaw**.
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## Streamlit Features

- **Predict:** Upload PNG/JPG or pick a row from dataset shards for inference.
  - **Threshold Tuning:** Confusion matrix, Precision, Recall, F1, ROC-AUC vs threshold.
  - **Validate (Stats):** Class balance and shard size distribution.
  - **Explore Dataset:** Browse sample rows and labels.
  - **Training Logs:** View training CSV logs and metrics.
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## Industrial Application

Integrates into QC pipelines to automate detection, reduce inspection time, and provide auditable statistics.