

# Week 13 - Project development

## Project Overview:

In this project, I have built an **Anomaly Detection System** using **Teachable Machine** and integrated it into a **Streamlit app**. The system is designed to detect anomalies in a **manufactured product**, similar to how an inspection system might be used to identify defects or irregularities in products during production.

## Steps Followed:

### 1. Teachable Machine Setup:

- I selected a **manufactured product** (different from the dataset used for InspectorsAlly) for this project.
- I uploaded images of the product to Teachable Machine's **Image Project** and trained the model with two classes: one for **Normal** (no anomaly) and one for **Anomalous** (with defects).
- After training, I exported the model from Teachable Machine as a **TensorFlow Lite model (.tflite)** and saved the class labels in a text file.

### 2. Streamlit App Integration:

- I then created a **Streamlit app** to integrate the trained model to perform anomaly detection.
- The app allows users to **upload an image** to check if the product is **Normal** or has an **Anomaly**.
- Additionally, I implemented a **live camera feed** functionality (bonus point) to allow real-time anomaly detection using the webcam, without the need to upload an image.

### 3. Model Optimization:

- I fine-tuned the model parameters, such as **Epochs**, **Batch Size**, and **Learning Rate**, to optimize the model's performance and accuracy in detecting anomalies.

### 4. Deployment:

- The final Streamlit app was deployed, and users can now use the system to detect product anomalies by uploading an image or using the camera feed in real-time.

## Streamlit App Link:

You can try the live application here: [Anomaly Detection System](#)