

Software Requirement Specifications
Automated PCB Defect Detection Using Deep Learning



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Meeting Details

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Summary

This Software Requirements Specification (SRS) describes the requirements for the project 'Automated PCB Defect Detection Using Deep Learning'. The project aims to automate the quality inspection process of Printed Circuit Boards (PCBs) by developing a deep learning–based system capable of identifying defects such as open circuits, shorts, mousebites, missing holes, spurs, and copper defects.

1. Introduction

Printed Circuit Boards (PCBs) are a core component in almost all electronic devices. During mass production, PCB defects such as shorts, opens, missing holes, copper excess, and spurs can occur, resulting in faulty devices, financial loss, and safety risks. Traditional PCB inspection is done manually or through rule-based vision systems that often fail to detect complex defects.

The **Automated PCB Defect Detection System** aims to replace human-dependent visual inspection with an intelligent deep learning system. This module uses YOLOv8 for defect detection and classification and serves as a building block for a fully automated Industry-4.0-based inspection pipeline. The system takes PCB images as input and highlights defects through bounding boxes and labels.

1.1 Purpose

The purpose of this project is to automate PCB quality inspection using a deep learning model capable of identifying multiple defect types. The system ensures fast, accurate, and reliable inspection to reduce human error and improve production efficiency.

1.2 Scope

Included in the system:

- I. Training a YOLOv8 model using the PCB Defects Dataset
- II. Detecting and classifying PCB defects (open, short, missing hole, mousebite, spur, copper)
- III. Preprocessing and splitting dataset into train/validation/test
- IV. Producing annotated output images with bounding boxes
- V. Visualizing predictions and generating evaluation metrics

Not included in this system (out of scope):

- I. Real-time camera integration
- II. Web or mobile deployment
- III. Multi-layer PCB defect detection
- IV. Cloud-based production-level deployment

1.3 Product Perspective

The system is a standalone module for automated image-based defect detection.

It can be integrated into:

- I. Industrial PCB production lines
- II. Automated optical inspection (AOI) machines
- III. Quality assurance software
- IV. Smart-manufacturing dashboards

1.4 User Characteristics

User Type	Characteristics
Quality Inspector	Basic understanding of PCB defects; uses system to view results
ML Engineer / Developer	Knowledge of Python, deep learning, and model training
Supervisor / QA Manager	Reviews detection reports and results
Student / Researcher	Uses system for academic learning

1.5 Similar Apps and Systems / Literature Review

Several PCB inspection systems exist, including:

- I. **Rule-based AOI systems** — rely on predefined image filters; struggle with unseen defect variations.
- II. **Traditional Machine Learning systems** — require hand-crafted features; limited adaptability.
- III. **Industrial AOI hardware** — expensive and inflexible for small-scale manufacturers.

Recent research emphasizes YOLO, Faster R-CNN, and ResNet architectures for PCB defect detection. YOLOv8 provides state-of-the-art speed and accuracy, making it ideal for real-time inspection scenarios.

1.6 Proposed Technologies

- I. **Programming Language:** Python
- II. **Deep Learning Framework:** PyTorch, Ultralytics YOLOv8
- III. **Data Processing:** OpenCV, NumPy, Pandas
- IV. **Visualization:** Matplotlib, Seaborn
- V. **Tools / IDE:** Jupyter Notebook, VS Code
- VI. **Dataset Source:** Kaggle PCB Defects Dataset

· 2. Requirements

The system will preprocess PCB images, train YOLOv8 using labeled defect examples, evaluate model performance, and detect defects on test images. It will produce bounding-box outputs, label predictions, and confidence scores.

The system will also generate evaluation metrics such as mAP, precision, recall, and confusion matrices.

2.1 Functional Requirements

FR001 – Dataset Preparation

Purpose: Prepare and convert PCB data into YOLOv8-compatible format.

User(s): ML Engineer

Input: Raw images, annotation labels

Process:

- Preprocess images
 - Convert annotations into YOLO format
 - Split data into train/val/test
- Output:** Ready-to-train dataset directory

FR002 – Model Training

Purpose: Train YOLOv8 model on PCB defects dataset.

User(s): ML Engineer

Input: Training dataset, hyperparameters

Process:

- Train model
 - Validate performance
 - Optimize hyperparameters
- Output:** Trained model weights

FR003 – Defect Detection

Purpose: Detect PCB defects in unseen images.

User(s): Quality Inspector

Input: New PCB image

Process:

- Run trained YOLO model
- Identify defects
- Draw bounding boxes

Output: Annotated image with defect labels and confidence scores

FR004 – Evaluation and Reporting

Purpose: Provide performance metrics and visual reports.

User(s): QA Manager, ML Engineer

Input: Validation and test datasets

Process:

- Generate confusion matrix
- Produce mAP, precision, recall, F1-score

Output: Charts, metrics, and performance reports

2.2 Non-Functional Requirements

Performance Requirements

- I. System must detect defects with high accuracy ($\text{mAP}@50 > 85\%$).
- II. Model inference time should be < 100 ms per image for GPU environments.

Reliability

- I. The system should perform consistently across diverse image types.
- II. Must handle corrupted or low-quality images gracefully.

Usability

- I. Interface must clearly display defect bounding boxes.
- II. Visual outputs should be easy for non-technical users to interpret.

Scalability

- I. Should allow integration with real-time AOI cameras in future.

Portability

- I. Must run on standard Python environments.

3. Use Cases and Flow of Processes

Below is the system-level use case diagram description.

3.1 Use Case 1: PCB Defect Detection

ID: UC001

Name: PCB Defect Detection

Description: This use case describes how the system identifies defects in PCB images.

Requirement(s): FR003

Actor(s): Quality Inspector

Precondition: Trained YOLOv8 model must be available

Postcondition: Annotated result image displayed

Basic Flow

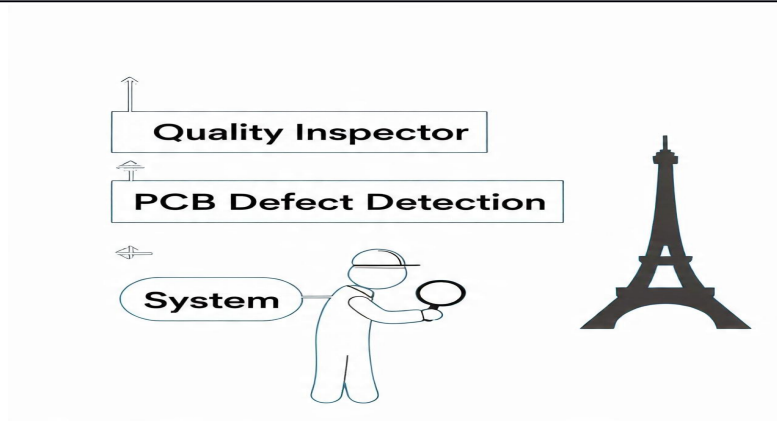
- I. User uploads/selects PCB image
- II. System loads trained model
- III. System processes image
- IV. YOLO model detects defects
- V. System displays image with bounding boxes and labels

Alternative Flow

- I. If model is not trained, system notifies user to train model first
- II. User may select multiple images for batch detection

Exceptions

- I. Invalid image format
- II. Model file missing or corrupted
- III. System unable to process due to memory issues



4. References

1. Kaggle PCB Defects Dataset
2. YOLOv8 Documentation
3. Industry 4.0 Research Papers
4. OpenCV Documentation
5. PyTorch Documentation