

## **Problem Definition & Design Thinking**

### **Title: Quality Control in Manufacturing**

#### **Problem Statement:**

In today's competitive industrial landscape, ensuring consistent product quality is vital for maintaining customer satisfaction and meeting regulatory requirements. Manufacturing processes often encounter variability due to human error, equipment malfunctions, raw material inconsistencies, and environmental factors. These issues can lead to defective products, increased costs, and damage to brand reputation.

The problem is how to implement a robust and proactive quality control system that can detect, predict, and minimize defects during the manufacturing process, rather than after production is completed.

#### **Target Audience:**

- Manufacturing companies aiming to reduce waste and improve quality.
- Quality assurance teams seeking to improve inspection and testing processes.
- Industrial engineers and operations managers.
- Companies adopting Industry 4.0 solutions for smart manufacturing.

#### **Objectives:**

- To design a quality control system that uses data-driven insights for real-time defect detection.
- To improve consistency and reduce variability in production processes.

- To incorporate automation and AI for predictive quality analytics.
- To ensure compliance with industry standards and regulations.

### **Design Thinking Approach:**

#### **Empathize:**

Workers on the production floor, quality inspectors, and engineers face daily challenges in identifying and resolving quality issues. Delayed detection can lead to batch-wide defects, recalls, and significant financial losses. It's essential to understand their workflow and the root causes of variability.

#### **Key User Concerns:**

- Downtime due to manual inspection.
- Lack of real-time defect alerts.
- Over-reliance on post-production quality checks.
- Difficulty in tracing defects to their origin.

#### **Define:**

The solution should enable continuous monitoring of manufacturing processes, leveraging sensors, machine learning, and real-time analytics. It must alert operators to anomalies and suggest corrective actions immediately.

#### **Key Features Required:**

- Sensor-based data collection from machines and processes.
- AI-powered defect detection and root cause analysis.
- Dashboard for real-time monitoring and alerts.
- Compliance and audit trail functionalities.

### **Ideate:**

Some potential ideas include:

- A machine vision system to inspect products automatically.
- Predictive maintenance tools using sensor data.
- A digital twin of the production line for simulation and quality optimization.
- Integration of statistical process control (SPC) with real-time feedback.

### **Brainstorming Results:**

- An AI system trained on past defect patterns to predict future quality issues.
- Visual analytics dashboard showing quality trends and KPIs.
- Mobile alerts for operators when process parameters deviate from the norm.

### **Prototype:**

Developing a prototype that includes:

- Sensor array and camera for capturing process and product data.
- Machine learning model for classifying and predicting defects.
- A user interface showing alerts, logs, and performance metrics.

### **Key Components of Prototype:**

- Industrial IoT sensors and image recognition hardware.
- Cloud-based platform for analytics and model training.
- User interface for process monitoring and corrective action logging.

### **Test:**

The prototype will be tested on a small-scale production line. Operators and quality managers will interact with the system, and their feedback will guide improvements.

### **Testing Goals:**

- Verify accuracy of defect detection and prediction.
- Assess ease of use and integration with existing workflows.
- Evaluate the reduction in defects and downtime.