



Optimizing the Vision-Guard AI models to speed up AI training process

UIIA

Vu Thanh Tung - s3963172 Dinh Xuan Minh - s3891847 Le Quang Dat - s3927251 Nguyen Linh Chi - s3878609 Le Tran Nhat Nam - s3911298

ACADEMIC SUPERVISOR

Mr. Pham Viet Hung

INDUSTRY SUPERVISOR

Mr. Nguyen Minh Chau Mr. Pham Cong Vo Mr. Pham Quang Vinh



SCAN ME

7

Background & Motivation

Manual visual inspection in semiconductor manufacturing creates quality risks through inspector fatigue, attention lapses, and subjective assessments. Al-driven inspection eliminates these human factors, providing consistent and objective quality control.

However, Al model development faces bottlenecks in dataset creation and performance validation. Traditional methods require extensive manual image collection, annotation, and time-intensive testing that delays deployment.

Synthetic Data

Generator

Model

Preparation

Model training

Z

Objectives

- Develop an Al-driven defect detection system for Intel's semiconductor production line.
- Improve dataset creation through synthetic generation and active learning to reduce cost and time.
- Train and fine-tune neural networks to achieve ≥95% accuracy while minimizing false positives/negatives.
- Deploy optimized models on edge devices to cut inference costs and ensure realtime performance.
- Build an interactive dashboard to visualize defect rates, confidence scores, and production metrics for decision-making.

7

Solution Design

Synthetic Data Generator

- Description: Generates additional training images from a small seed dataset.
- Purpose: Automates dataset expansion by replicating original product images and adding realistic defects (e.g., foreign material on pusher, victim scratches). This speeds up dataset preparation and reduces the need for manual photography of every product.

Confusion Matrix Testing

- Description: A validation tool for testing Al models after training.
- Purpose: Instead of manually testing products on Intel Geti one by one, employees can
 upload a dataset, and the app will automatically display AI predictions against actual
 product types. Users mark pass or fail, and upon completion, the app exports an Excel
 file containing images, results, and a confusion matrix score.

Al Creek 2.0

- Description: A station-level web application integrated with the Intel Geti SDK.
- Purpose: Allows employees to capture images directly through the web browser, where Al generates predictions. Operators decide pass or fail, while the dashboard displays key production metrics for each product. Supervisors can access logs, including operator names, IDs, and timestamps, to ensure accountability and traceability.



Confusion Matrix Testing



Model



Al Creek 2.0



Predicted





Metc SAM 2



Production line Station PC 1..N History queries Products on conveyor Al Creek Web Defect Detection Model GPU inference Read time results display Orchestrator Read/Write Read/Write Result Store / Logs (DB,NAS)

Future works

Future directions:

- Complete fine-tuning for all product models and finalize confusion matrix testing.
- Improve robustness against factory environmental noise (lighting, vibration, dust).
- Deploy the system in Intel's production line for large-scale testing.
- Enhance the dashboard with real-time CMMS data integration.
- Explore hardware upgrades and advanced ensemble methods to further increase reliability.

Results

- Developed Dataset Generator Application for dataset creation automation.
- Built Al Creek application for displaying dashboard, showing metrics and graphs from CMMS data.
- Trained and optimized product models, increasing confusion matrix score (Insert specific metrics).



46/50 sample correct (92%) 35/50 sample correct (70%)

