

PWS: Potential Wafermap Scratch Defect Pattern Recognition with Machine Learning Techniques

Peter Yi-Yu Liao – Assembly & Test Kaohsiung, NXP Semiconductor Taiwan

Ken cc Cheng – Assembly & Test Kaohsiung, NXP Semiconductor Taiwan

Leon Chou - Assembly & Test Kaohsiung, NXP Semiconductor Taiwan

Andrew Yi-an Huang – Assembly & Test Kaohsiung, NXP Semiconductor Taiwan

Katherine Shu-Min Li – Dept. of Computer Science and Engineering, National Sun Yat-sen University, Taiwan

Sying-Jyan Wang – Dept. of Computer Science and Engineering, National Chung-Hsing University, Taiwan

Gus Chang-Hung Han – Dept. of Computer Science and Engineering, National Chung-Hsing University, Taiwan



Abstract

- Introduction
- Problem Definition
- Methods
- Experiment Results
- Conclusion

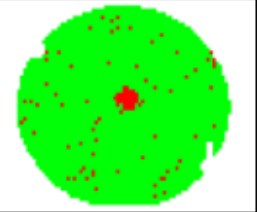
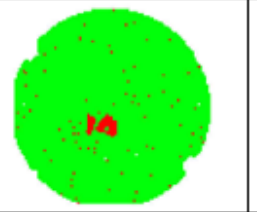
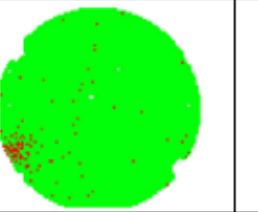
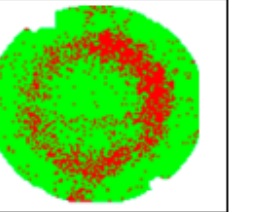
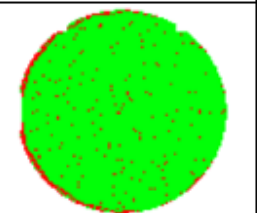
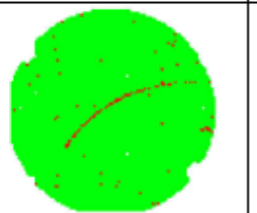
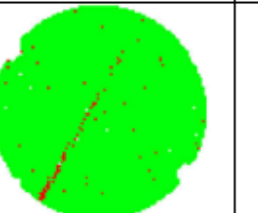
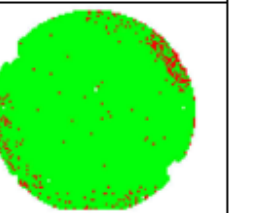
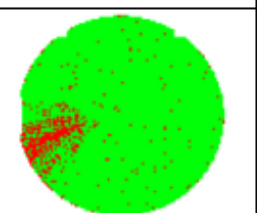
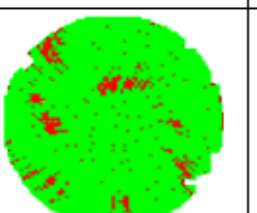
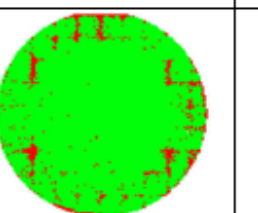
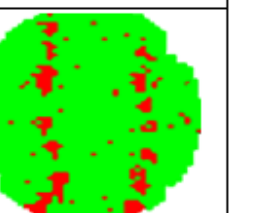
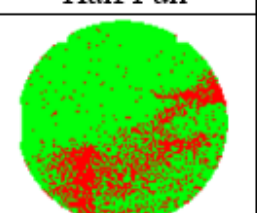
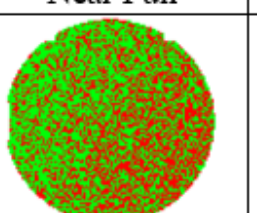
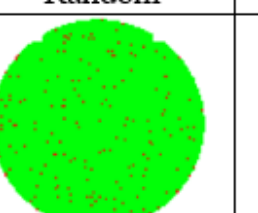
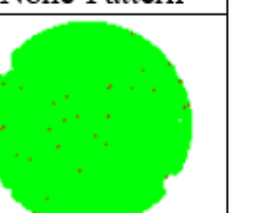


Introduction

- Wafer Defect Maps in wafer test show defects in fabrication process and test process
- Wafer defect patterns are the result of various problems in the manufacturing and wafer test process
- The Defect Patterns and Failure Modes of defects in a wafer are highly correlated to the Root Causes
- Through Wafer defect pattern recognition we can conduct Root Cause Analysis to identify possible problems in the fabrication and wafer test process



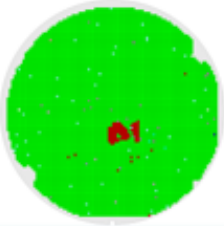
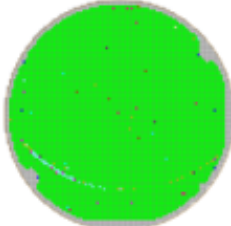

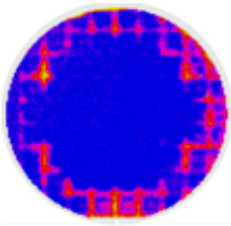

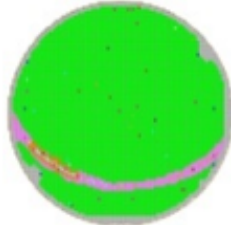
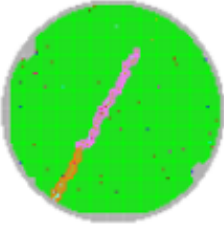
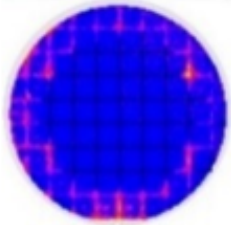
Introduction

| Center | Local | Edge Local | Donut |
|---|--|---|---|
|  |  |  |  |
| Edge Ring | Scratch | Line | Tennis |
|  |  |  |  |
| Sector | Radiation | Grid | Ladder |
|  |  |  |  |
| Half Full | Near Full | Random | None-Pattern |
|  |  |  |  |



Problem Definition

- The potential defect are the discontinuity in spatial defect patterns corresponds to dies passing wafer test, they are caused by variation of mechanical force
- Since they fall inside a given defect pattern, it implies that such dies may suffer from the same cause that is responsible for the defect pattern

| WDD | Local | Scratch | Line | Special: Matrix |
|------------------|---|--|--|--|
| Conventional WDD |  |  |  |  |
| Potential WDD |  |  |  |  |



Problem Definition

- Scratch and Line types of wafer patterns are often the most obscure
- Potential Wafer Scratch pattern recognition system
- Identify incomplete scratch pattern
- Reduce the classification effectiveness in wafer defect diagnosis
- Improve the prediction/classification accuracy



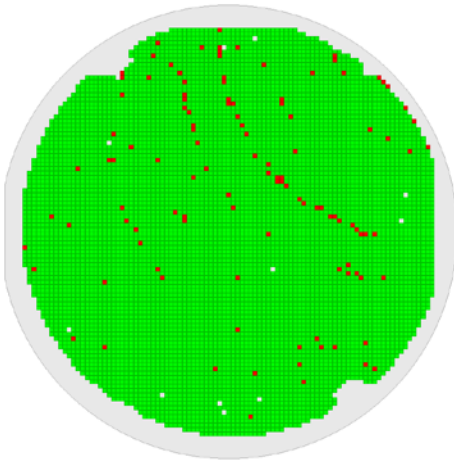
Methods

- Phase I: Wafer Lot Test
- Phase II: Wafer Clustering
- Phase III: Intra-Cluster Point Filling
- Phase IV: Inter-Cluster Point Filling
- Phase V: Machine Learning Based Classification

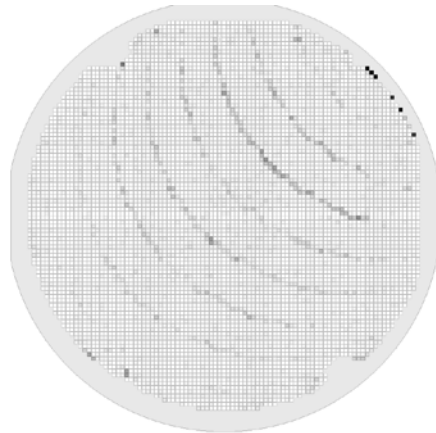


Wafer Lot Test

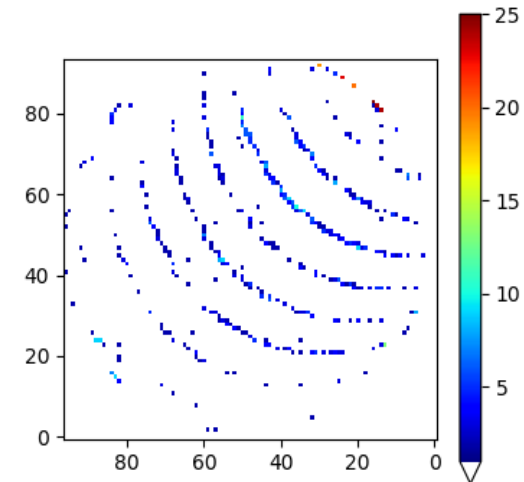
- Wafers are processed in batches of wafer lots in manufacturing process
- Search for potential scratch pattern in wafer maps per lot
- Stack all wafers in same lot into one lot wafer map



Single wafer



Lot wafers



Wafer Clustering

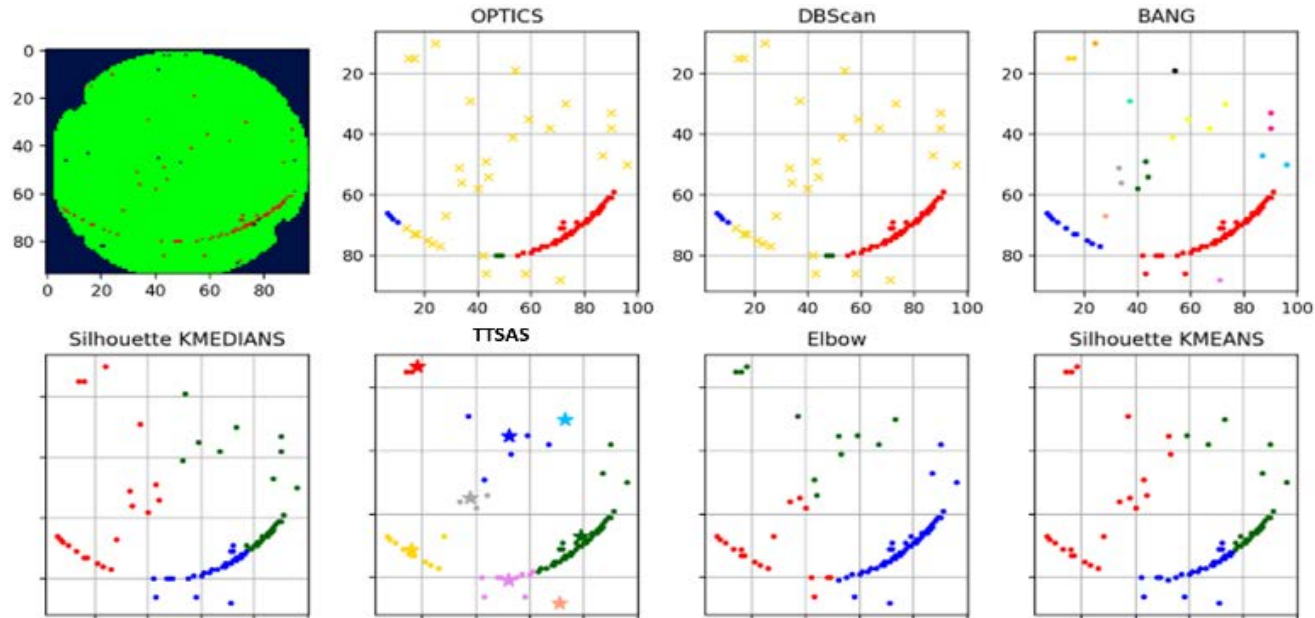
- Defective dies in a wafer may be caused by various problems, and defects caused by the same problem tend to form a local cluster
- Clustering algorithms
 - OPTICS (Ordering Points To Identify the Clustering Structure)
 - DBSCAN (Density-based spatial clustering of applications with noise)
 - TTSA (Two-Threshold Sequential Algorithm)
 - Elbow
 - BANG
 - K-Medians
 - K-Means



Wafer Clustering

- Rand Index (RI)

- $RI = \frac{\#TP + \#TN}{\#TP + \#FP + \#FN + \#TN}$



Wafer Clustering

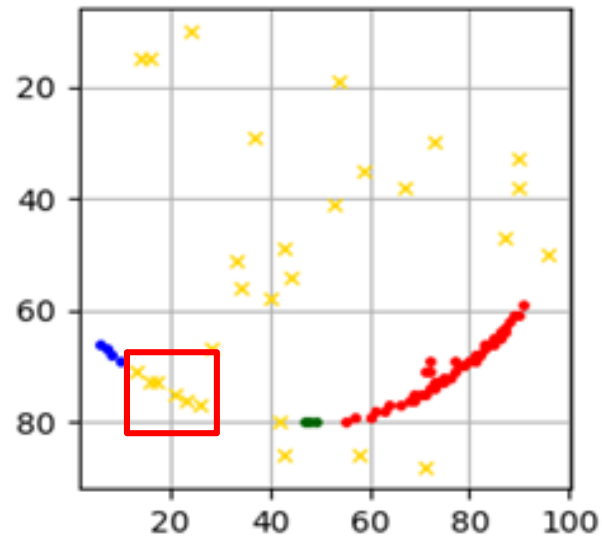
- OPTICS

- Based on defect density to cluster defective dies of a defect type
- A generalization of DBSCAN, and can solve one of DBSCAN's major shortcomings: effective clustering over density-variant data set
- Robust against variation of shapes and sizes in Scratch pattern



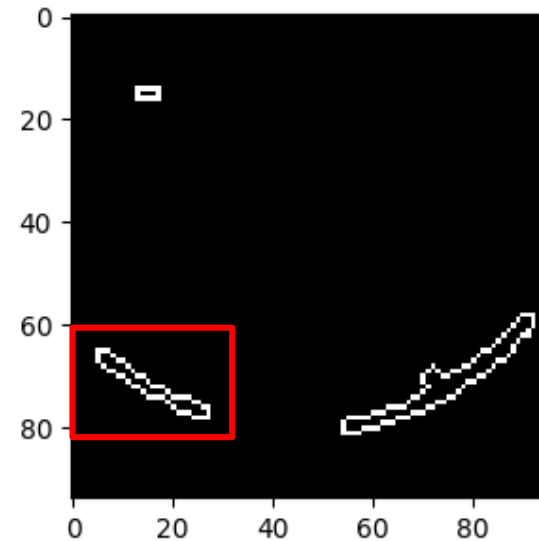
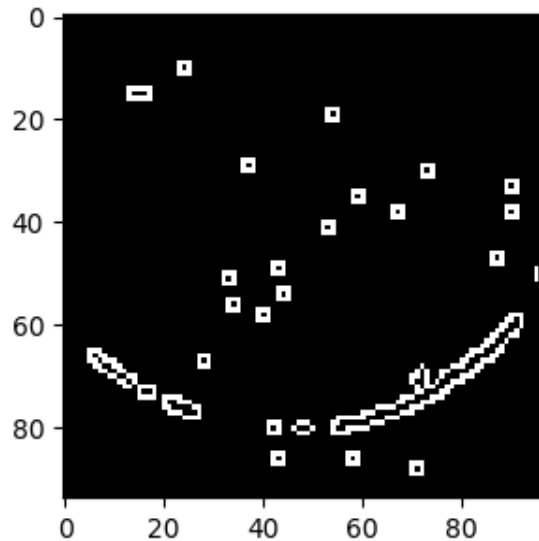
Intra-Cluster Point Filling

- The clustering process removes the random defect dies not in the pattern, but sometimes can also remove die that belong to the pattern but not obvious
- Scattered points are not put in any cluster as the defect density around these points is lower than the threshold.



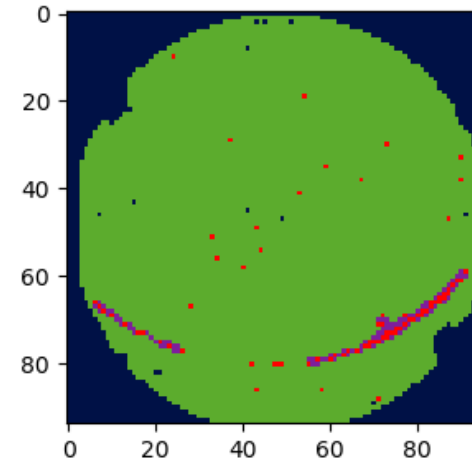
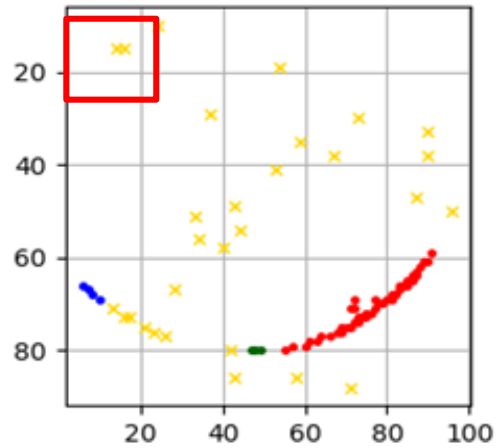
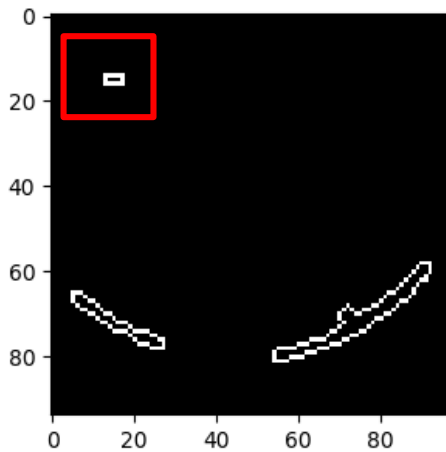
Intra-Cluster Point Filling

- Canny Edge Detection
- Region Labeling & Filtering
- Image closing operation



Intra-Cluster Point Filling

- Combine the cluster result and the image process result
- Flood Fill Algorithm



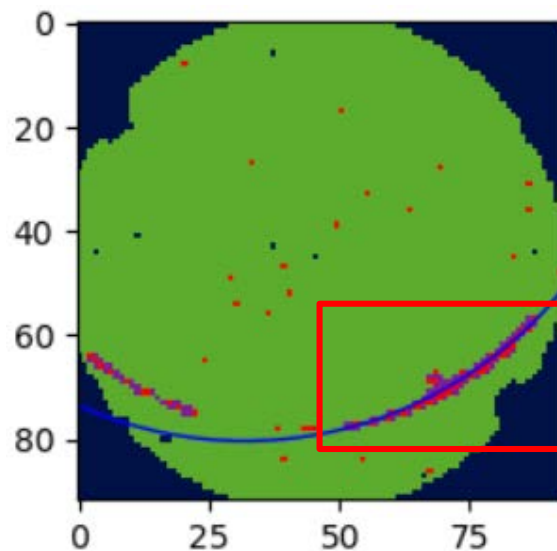
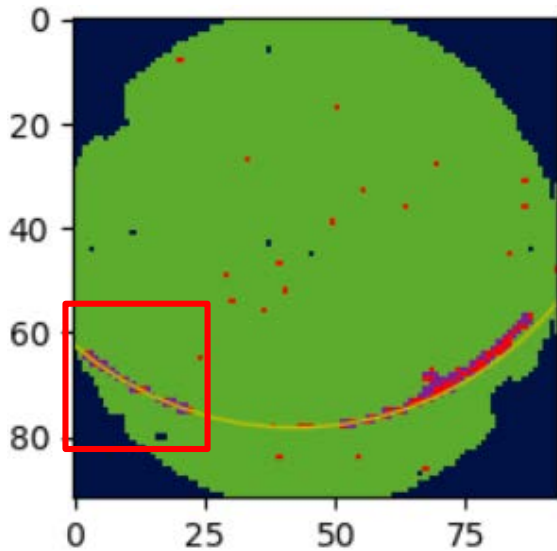
Inter-Cluster Point Filling

- For different clusters with potential Scratch, we need to judge:
 - 1. **Integrated Scratch:**
 - There are disconnected points in Scratch, but with the same arc shape.
 - 2. **Individual Scratch:**
 - Due to different arc shapes, they belongs to individual scratches with their own features



Inter-Cluster Point Filling

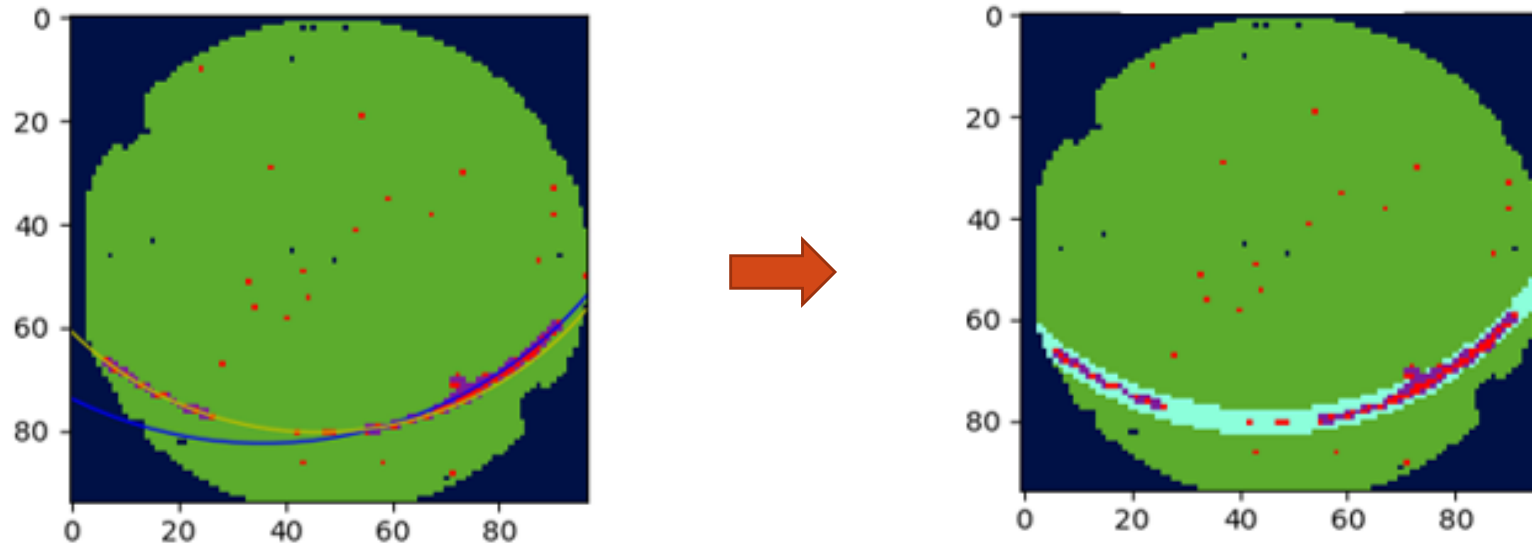
- Scratch Generation
 - Generate circle that matches the shape of each cluster



Inter-Cluster Point Filling

- Integrated *vs.* individual Scratches
 - Judge whether the intersection of individual scratches exists
 - Jaccard Index (Jaccard similarity coefficient) is used to compare similarity between two arc shape

$$JI(A, B) = \frac{|A \cap B|}{|A \cup B|} = \frac{|A \cap B|}{|A| + |B| - |A \cap B|}$$



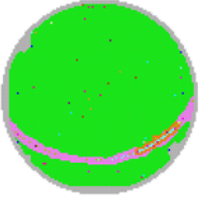
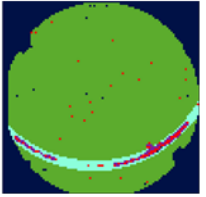
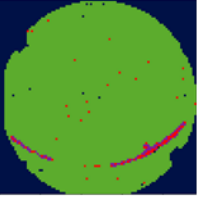
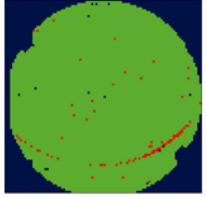
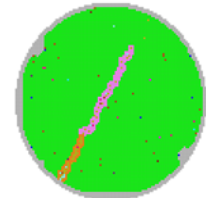
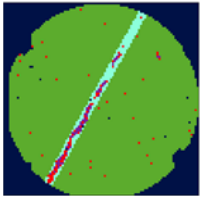
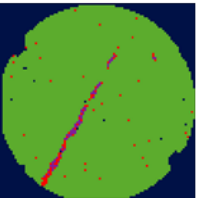
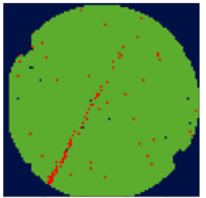
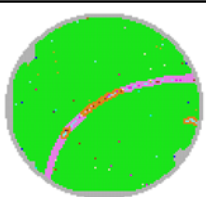
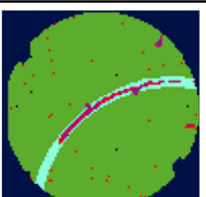
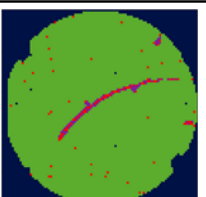
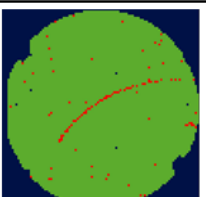
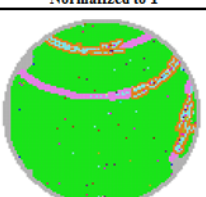
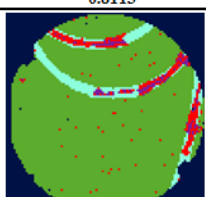
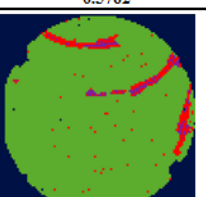
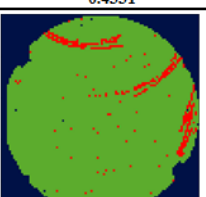
Machine Learning Based Classification

- Ada Boost and Decision Tree method
 - Spatial Feature
 - Geometry-based Feature
 - Density-based Feature
 - Radon-based Feature



Experiment Results

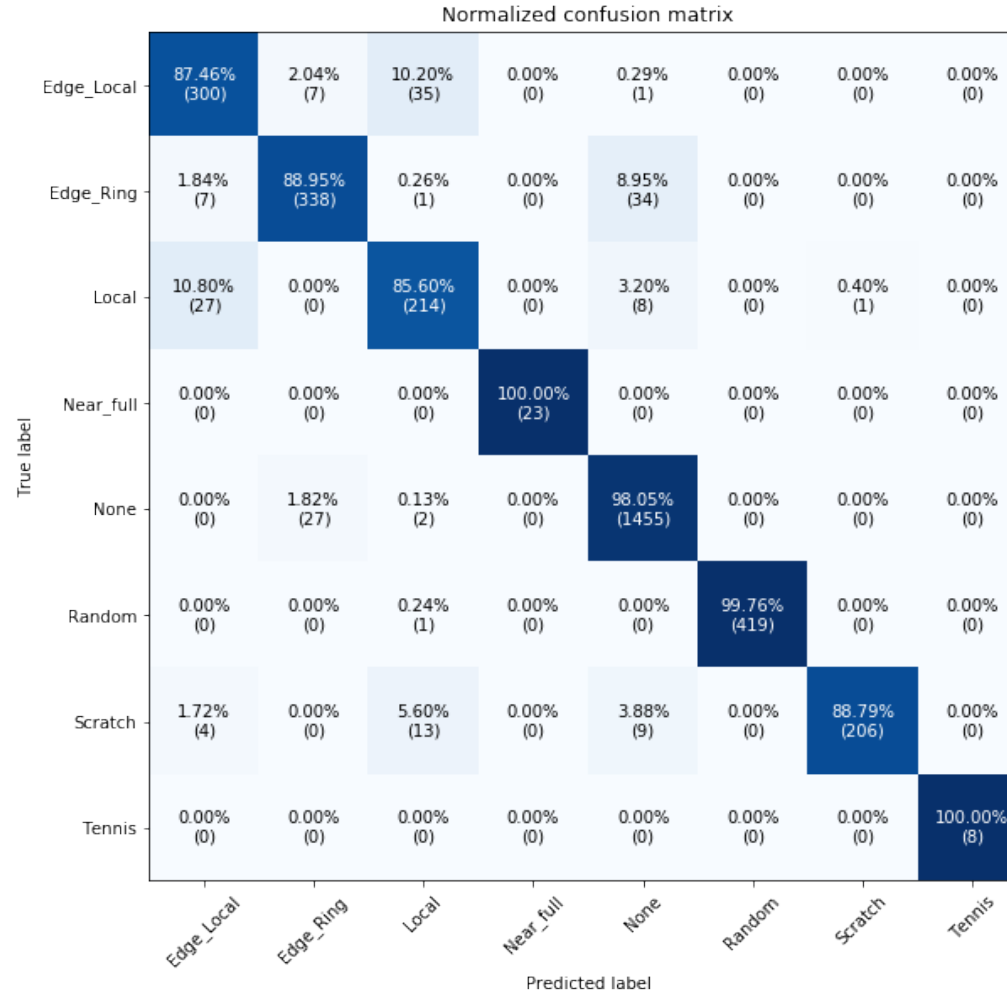
- Similarity ranking

| Ideal/Manual Edited wafermap | Intra + Inter Clustering | Intra Clustering | Wafermap-under-Test (WUT) |
|---|--|---|---|
|  |  |  |  |
| Normalized to 1 | 0.8478 | 0.5073 | 0.3511 |
|  |  |  |  |
| Normalized to 1 | 0.8242 | 0.6277 | 0.4612 |
|  |  |  |  |
| Normalized to 1 | 0.8113 | 0.5762 | 0.4331 |
|  |  |  |  |
| Normalized to 1 | 0.6955 | 0.632 | 0.5358 |



Experiment Results

- 8 Defect patterns
- Total 3140 wafers
- 80% training, 20% test



Conclusion

- An automatic defect pattern recognition system demonstrated to search for potential scratch/line defect types
- Experimental results show the proposed method can achieve more than 89% prediction accuracy for scratch/line types, and higher than 94% for all common wafer defect types
- The quality and reliability of products can be significantly improved and cost of final test can be reduced.



THANK YOU

