

Wafer Defect Classification By Deep Learning

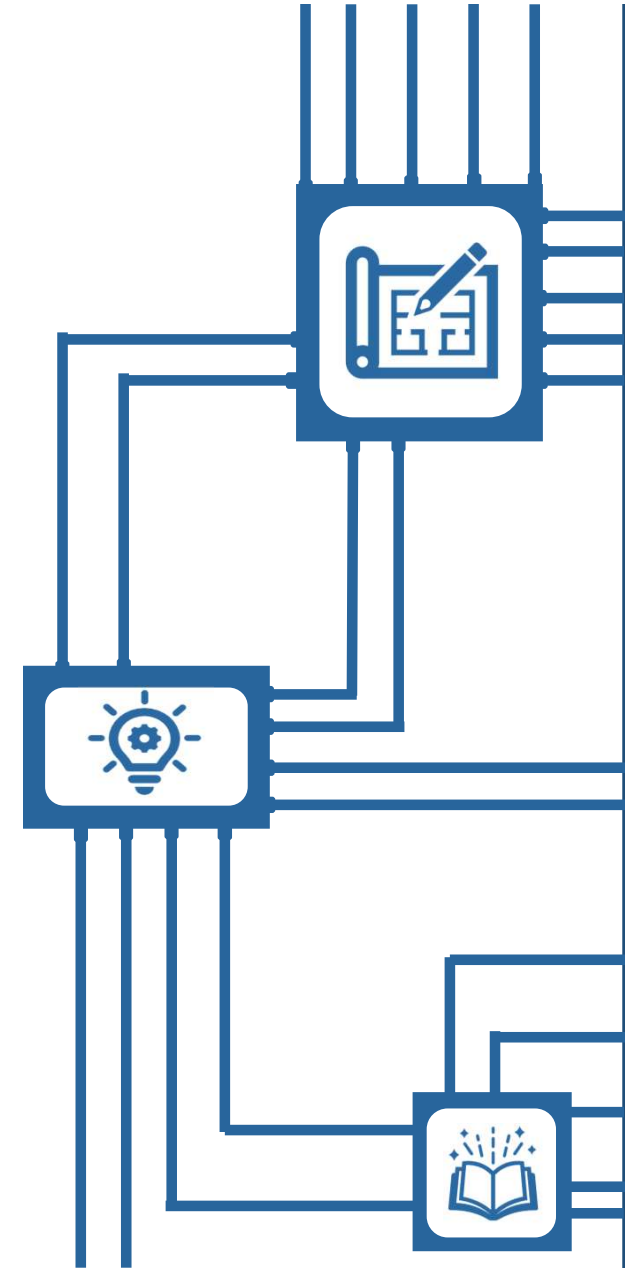
Capstone Design in Industrial Engineering



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Problem Description

- ✓ The most important part in semiconductor manufacturing is the “YIELD”($\text{Normal DIE} / \text{Total DIE}$)
- ✓ A lot of work is being done to increase the yield to improve profitability.
- ✓ Technology that predicts defects before final production of semiconductors is becoming more important nowadays.
- ✓ In this project, We built a CNN Modeling to enable classify and prediction of the causes of wafer defects by using Deep Learning(CNN Modeling).



반도체 수급 좋아진다더니
더 늘어난 현대차 신차 대기
... “싼타페 HEV 20개월”

(02/Sep/22_Hankook Ilbo)



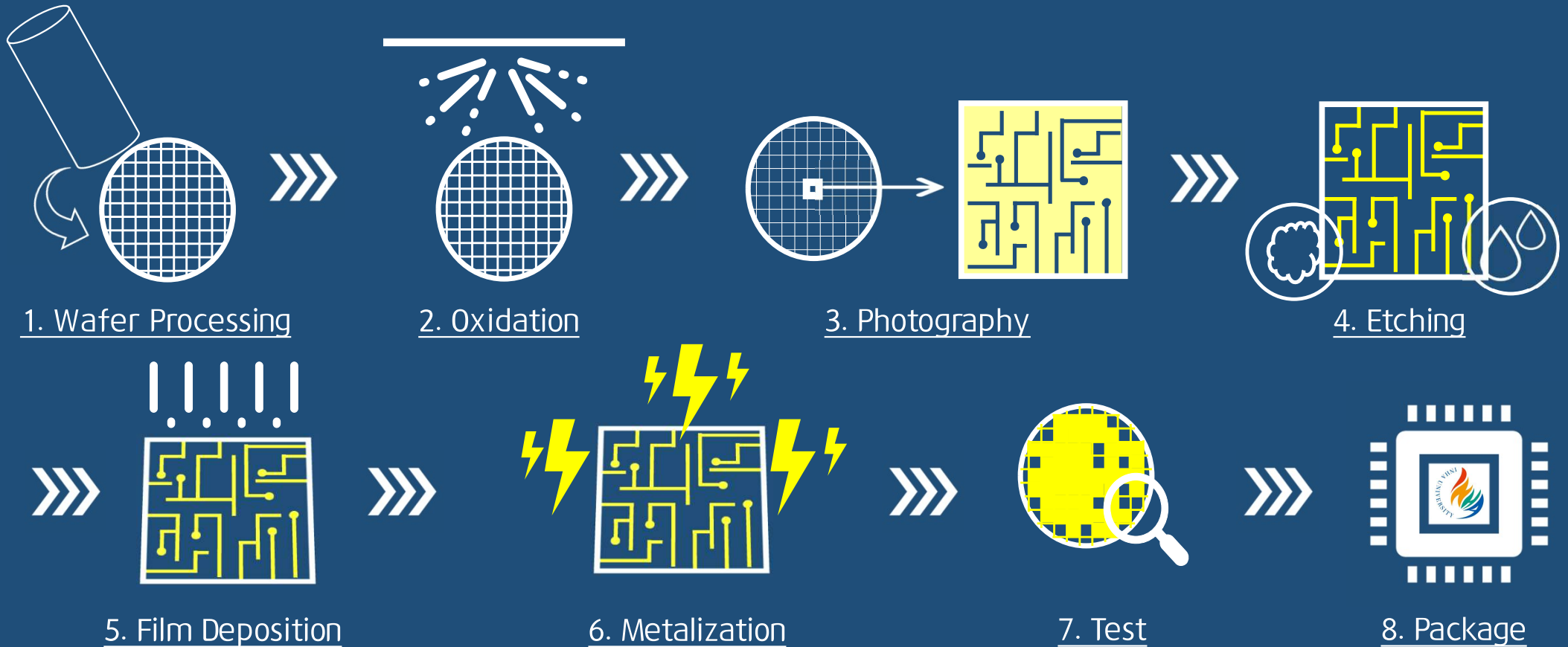
삼성전자 3나노 수율 탓에
양산 지연, 올해 자체 반도체만
적용 가능성

(14/Apr/22_Business Post)



What is Semiconductor Process?

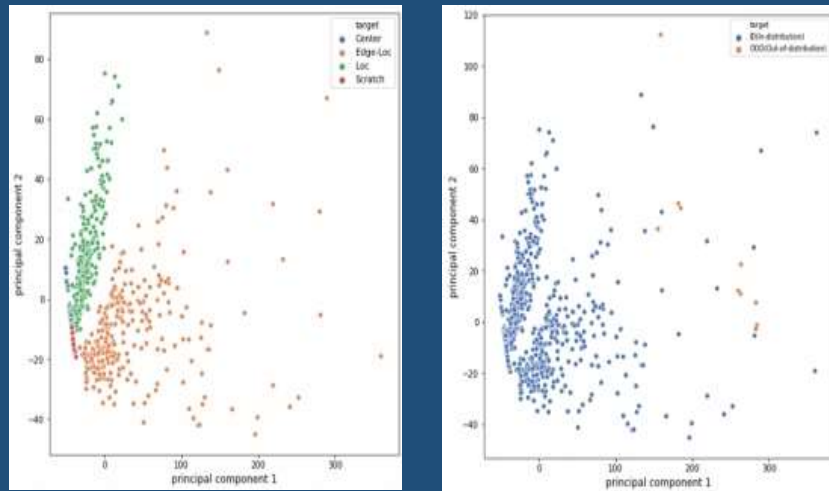
✓ For manufacturing of each semiconductor products, there are processes divided into 8 steps



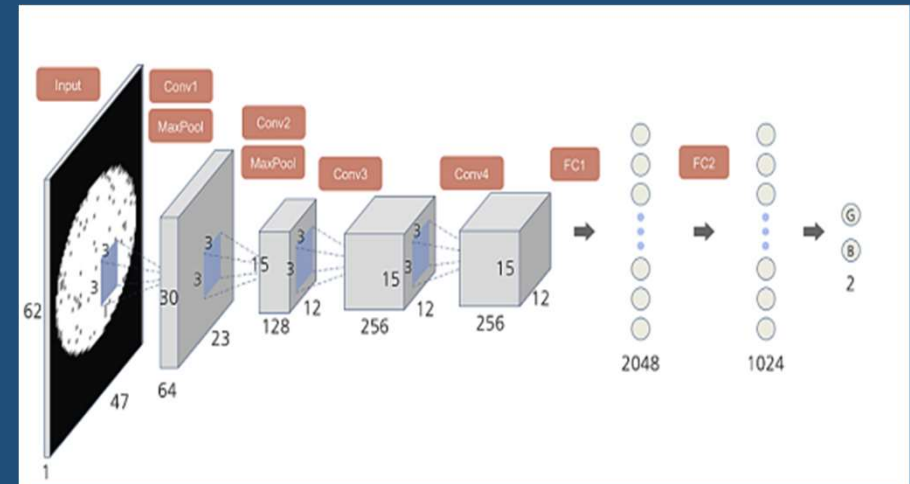


Prior Technology & Reference

A Multi-Class Data Description Based Method for Classifying Predefined Defect Patterns and Detecting New Defect Patterns of Wafer Bin Maps



Wafer Map-based Defect Detection Using Convolutional Neural Networks



811457 rows x 6 columns



Data Analysis & Preprocessing

- ✓ In the semiconductor industry, 1 lot usually means 25 sheets. The lots used in the project are 47,543 [47,543(lots) * 25 = 1,188,575(sheets)]
- ✓ Therefore 1,188,575 indexing should occur, but the total number is about 811k. So deleted the lotName and waferIndex.

	waferMap	dieSize	lotName	waferIndex
0	[[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ...	1683.0	lot1	1.0
811456	[[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 2, 1, 1, ...	600.0	lot47543	2.0
811457 rows × 6 columns				

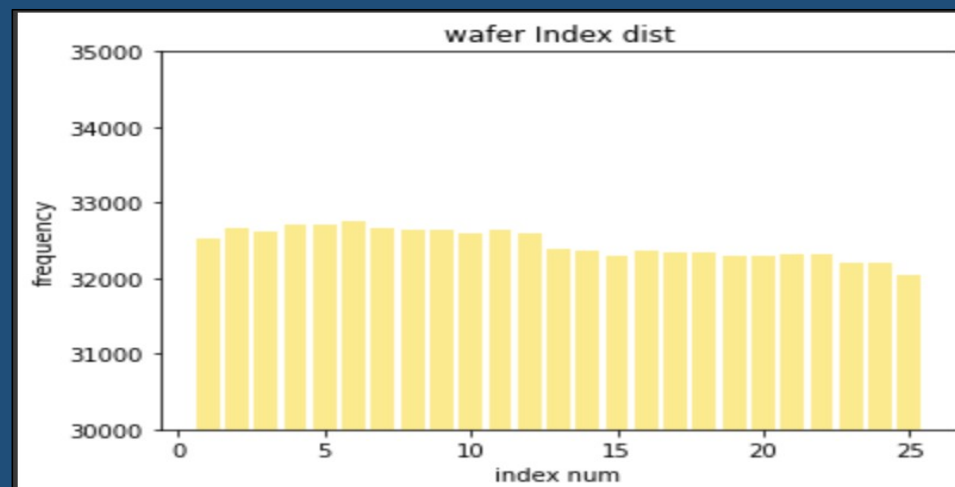


Figure 2. Wafer Index dist



Data Analysis & Preprocessing

※ Result ※

Defect Type	Num
C e n t e r	90
D o n u t	1
E d g e - L o c	296
E d g e - R i n g	31
L o c	297
N e a r - f u l l	16
R a n d o m	74
S c r a t c h	72
N o n e	13,489

※ Type of Wafer Defect ※

0

»»» Blank

1

»»» Normal

2

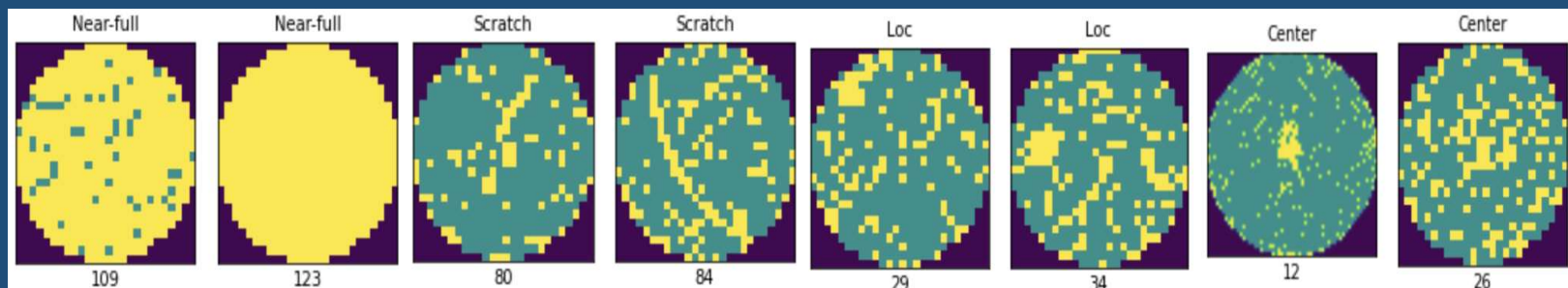
»»» Defect

Near-full : 16

Scratch : 72

Loc : 297

Center : 90

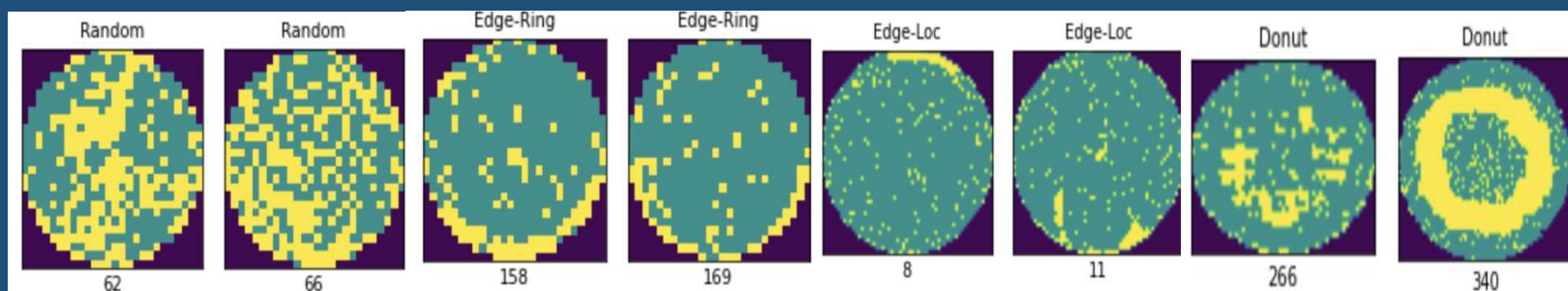


Random : 74

Edge-Ring : 31

Edge-Loc : 296

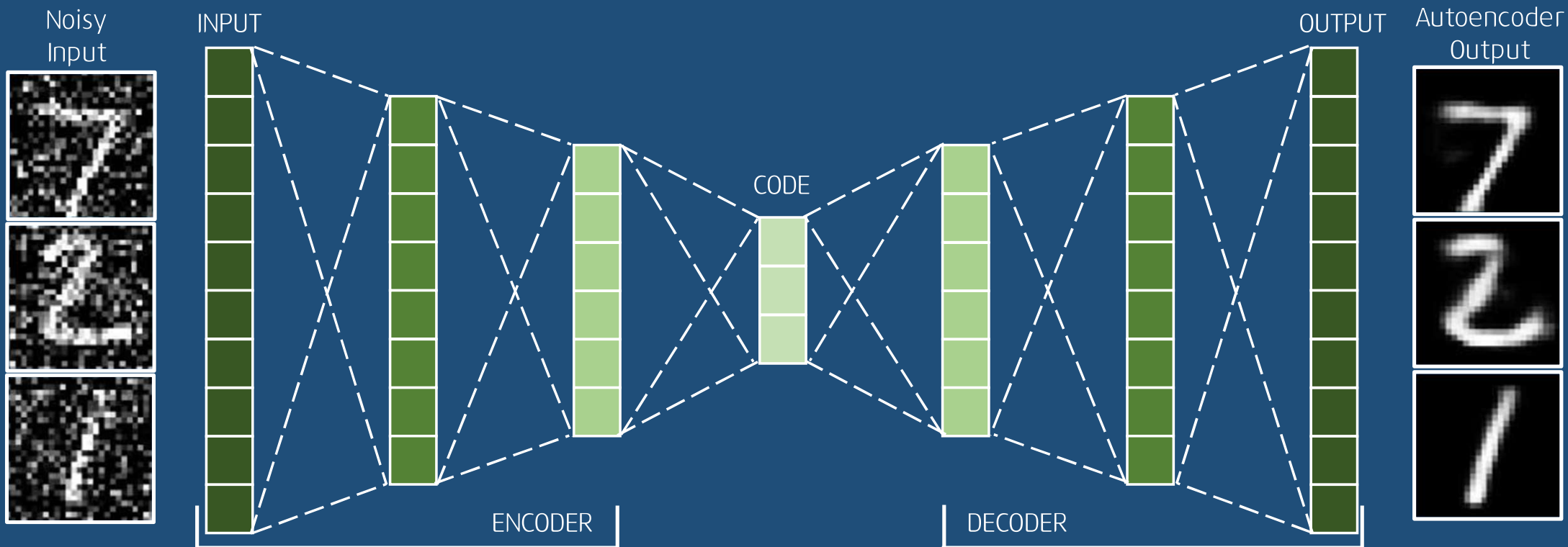
Donut : 1





Data Analysis & Preprocessing

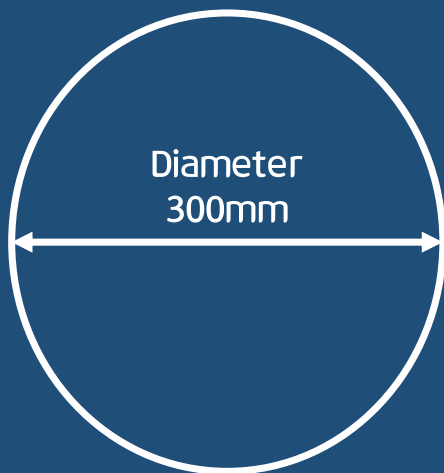
- ✓ For data augmentation, artificially added noise to the high-quality image and proceeded with encoder / decoder process.



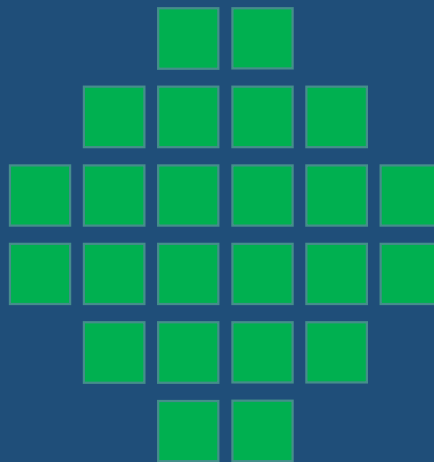


Data Analysis & Preprocessing

- ✓ From an economic point of view, Net Die is important because it is directly related to the Wafer unit price.
- ✓ The number of Net Die is Wafer Area / Die-Size, Therefore net die is inversely proportional to the Die-Size. (Die-Size increases as it gets smaller)



Wafer Area = 70,659mm²



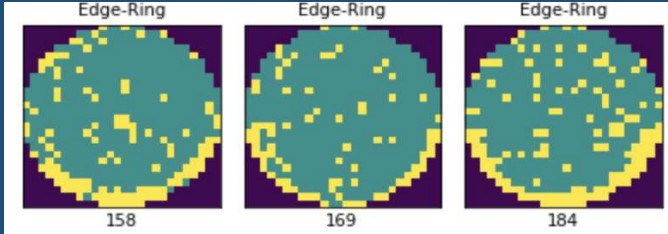
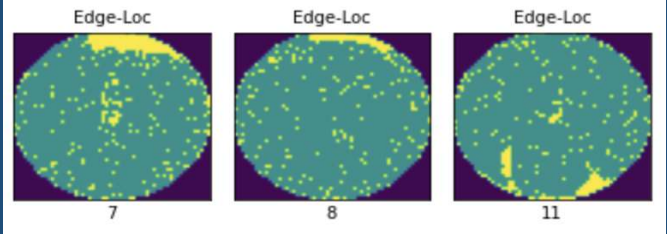
number of Net Die = 70,659mm² / \mathcal{X} → Die-Size

Die-Size	number of Net Die (\mathcal{X})
130mm ²	543개
100mm ²	706개



Data Analysis & Preprocessing

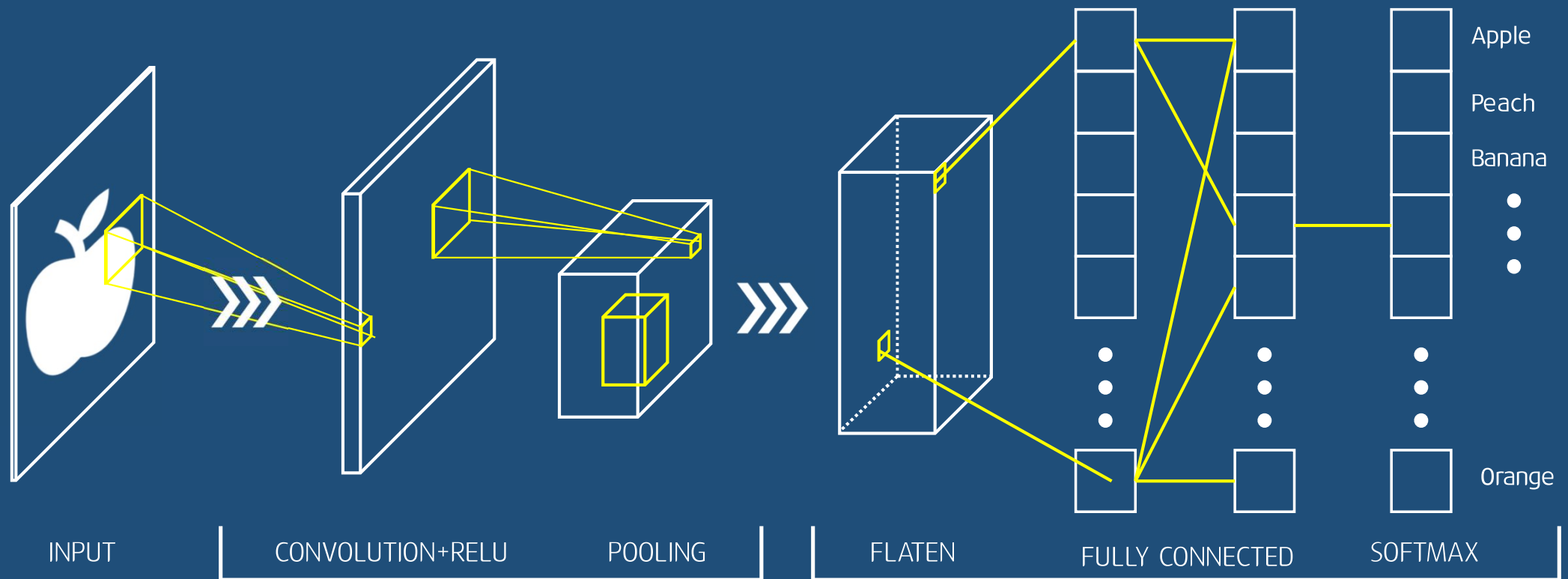
- ✓ In checking Net Die ($\text{Normal Die Size} / \text{Wafer Size}$) to ensure that all defect types are critical, We found that large numbers of abnormal chips are not always valid.
- ✓ Edge-log and edge-ring defects showed higher values than other types of Net Die, So focused on that type and performed augmented learning by using autoencoder.

	Edge-Ring	Edge-Loc
Image of defect		
Net Die	4.00	3.98
Before Autoencoder	31	296
After Autoencoder	7,037	7,400



Modeling

✓ We used CNN Modeling Construction that One of the type of deep learning model.





Evaluation

※ Performance Results ※

Number of Bad Classes & Whether it is augmented	Forecast Accuracy
Bad Class 2 / Argumentation X	90%
Bad Class 2 / Argumentation ○ : 3,000EA	89%
Bad Class 2 / Argumentation ○ : 5,000EA	92%
Bad Class 2 / Argumentation ○ : 7,000EA	93%

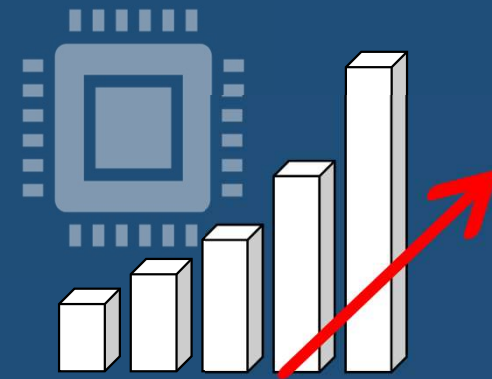


Solution

- ✓ The prediction accuracy of the model can be trusted by resolving the imbalance of target classes that occur in actual industrial sites.
- ✓ It is possible to improve the yield by detecting wafer defect patterns before the process is finished and request an engineer or department in charge to prevent defects in advance.



Balanced change in target class



Increased yield through pre-detection



The image features a central white square with a dark grey border, representing a microchip. Inside this square, the words "Thank You!" are written in a bold, black, sans-serif font. The chip is surrounded by a dark grey rectangular frame. Four vertical lines extend from the top and bottom of the frame, each ending in a small dark grey circle. The entire chip is centered within a large white circle. The background is a solid dark blue color, with several thin, dark grey lines extending from the left and right sides, suggesting a circuit board or network. The lines are arranged in a way that they appear to be connected to the chip, with some lines extending further outwards.

**Thank
You!**