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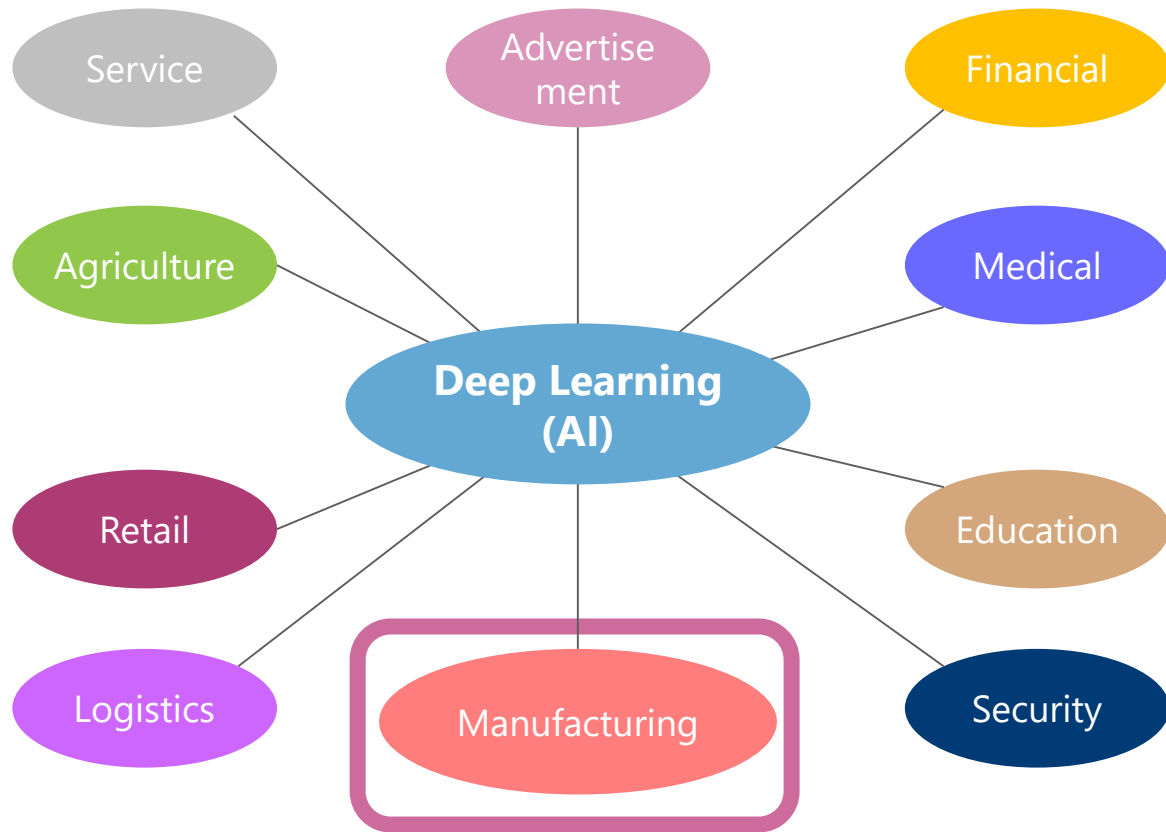
## Practical Application of Deep Learning in Smart Factory : Visual Inspection System of Semiconductor Laser

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Fujikura Ltd. and \*Optenergy, Inc. (Fujikura group)

# Today's Topic: Deep Learning in Manufacturing



## Difficulty of DL in Manufacturing

- Few defective images.
- The requirement is quite different from ordinary DL.  
ex. image size, criteria of classification, etc.

We have successfully overcome these.

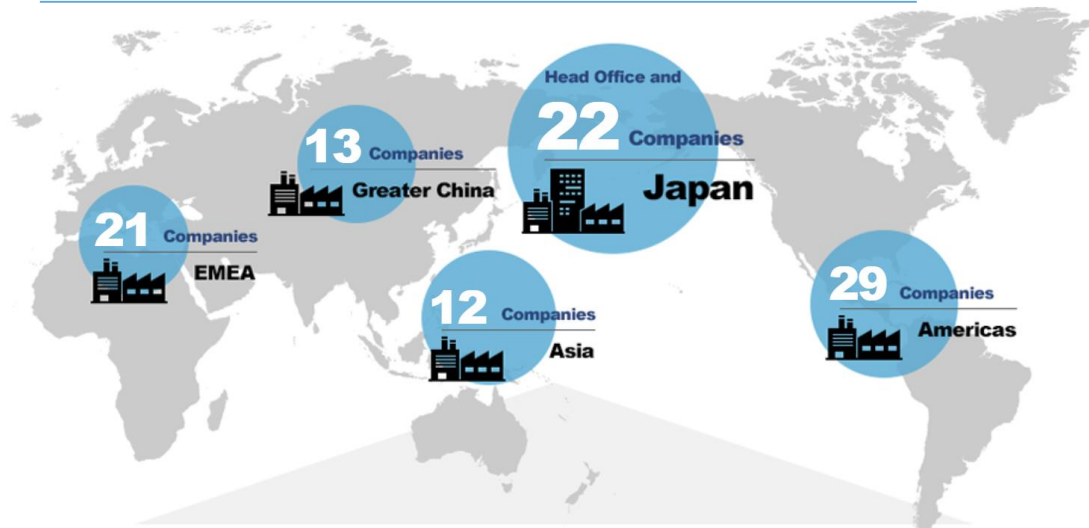
- The platform of DL visual inspection system was developed to apply the various production line.
- A visual inspection system has been implemented to actual production line.

1. Introduction of our company
2. Fujikura's "Monodukuri innovation"
3. Fiber Laser
4. Visual inspection using deep learning
5. Other application, future work.

# Fujikura Ltd. corporate profile

- Headquarters  
Tokyo, JAPAN
- Founded  
February 1885 (132 Years)

- Head office and consolidated companies



**Fujikura Group: 29 countries, about 140 companies**

- 4 business Areas



## Power & Telecommunication Systems Company

Optical Fiber / Optical Fiber Cable / Equipment & Components/Network Device / Optical Parts / Communication Cable / Power Cables/Industrial wires / Accessories /Bare wires / Aluminium wires / Enameled Wires



## Electronics Business Company

FPC / Connector / Electronic Wire / HDD Parts / Thermal Product / Sensor



## Automotive Products Company

Wire Harness / Automotive related Components



## Real Estate Business Company

Fukagawa GATHARIA

# History of our company

Fujikura has

**130** years history

**1885**

Establishment

**1923**

Restoration after Earthquake and  
development of Technology

**1970**

Oil Shock and Fluctuating Market

**1890**

Starting Operation

**1945**

From Post-war Recovery  
to High Economic Growth

**1991**

Globalization Diversification

**Now**

**Second Industrial Revolution**

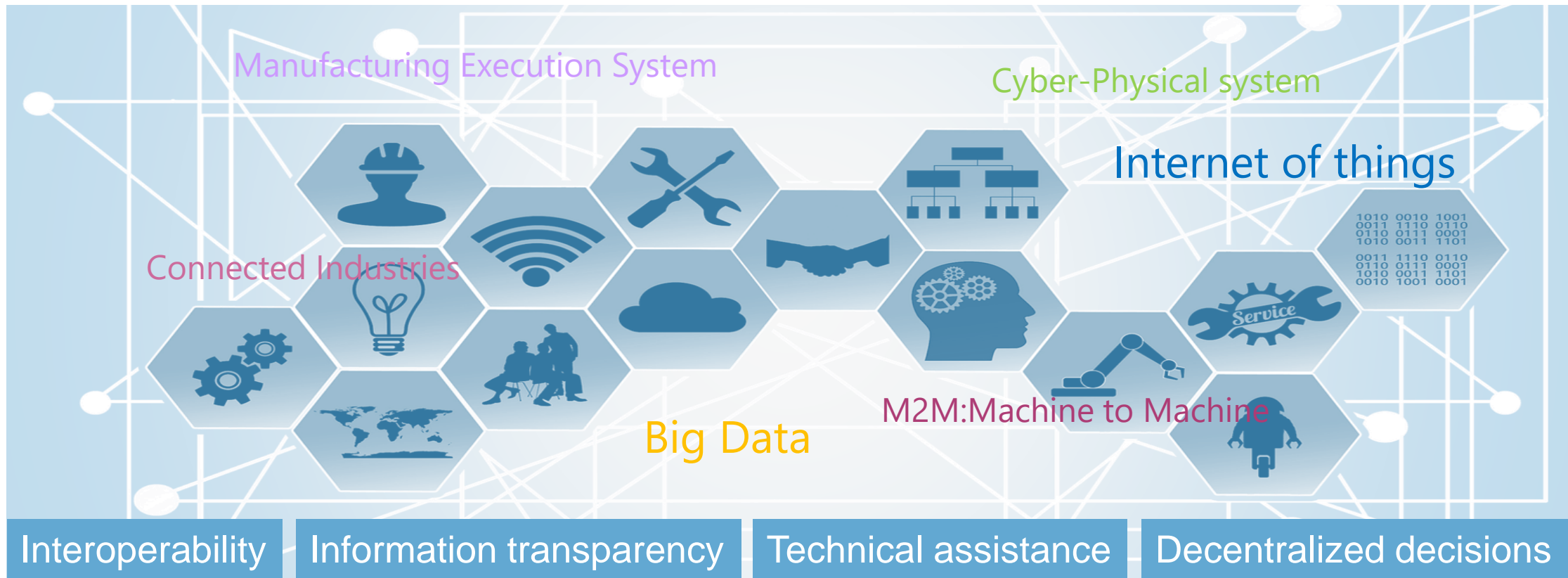
Electrification, Mass Production, ...

**Third Industrial Revolution**

Computer, Robot, Automation ...

**Toward  
Forth  
Industrial  
Revolution  
(Industrial 4.0)**

# Industry 4.0 & Smart factory

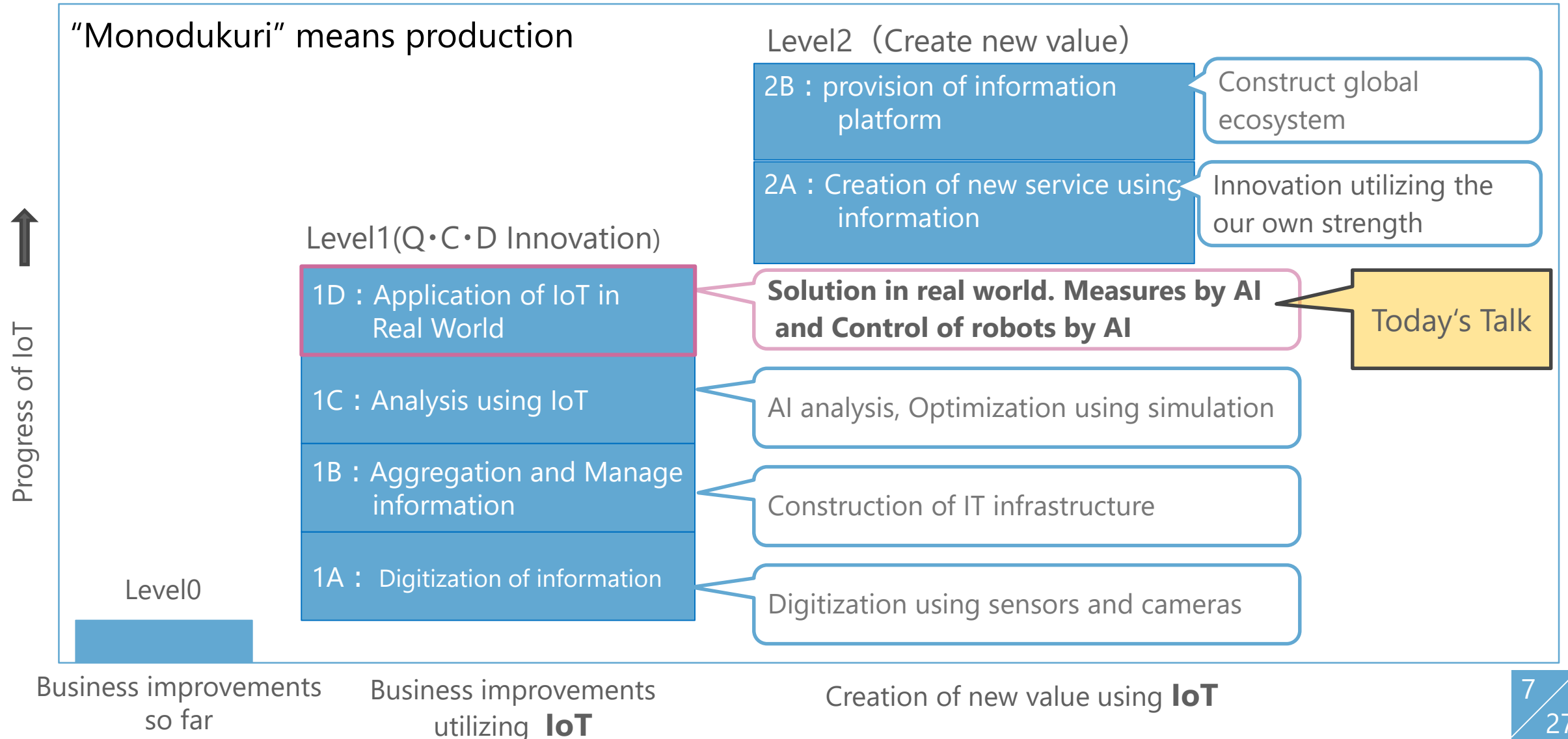


- Highly flexible mass production.
- Condition awareness of machine and process
- Intelligent support of workers
- Product quality
- Just-in-time maintenance and near-zero downtime

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## 2.1 “Monodukuri Innovation” in **Fujikura**

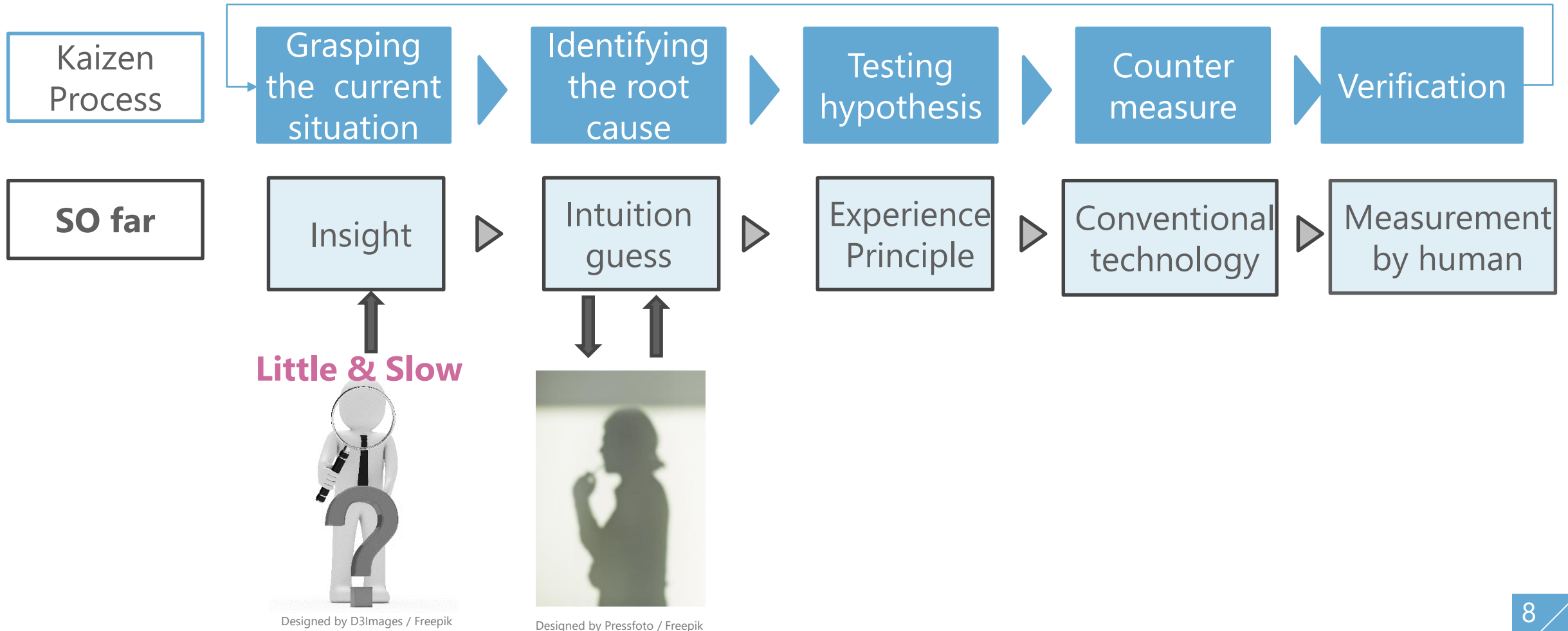
- Our main business field is Level 1:Q•C•D Innovation, Level 2A:Creation of new service





## 2.2 “Monodukuri Innovation”(Level 0)

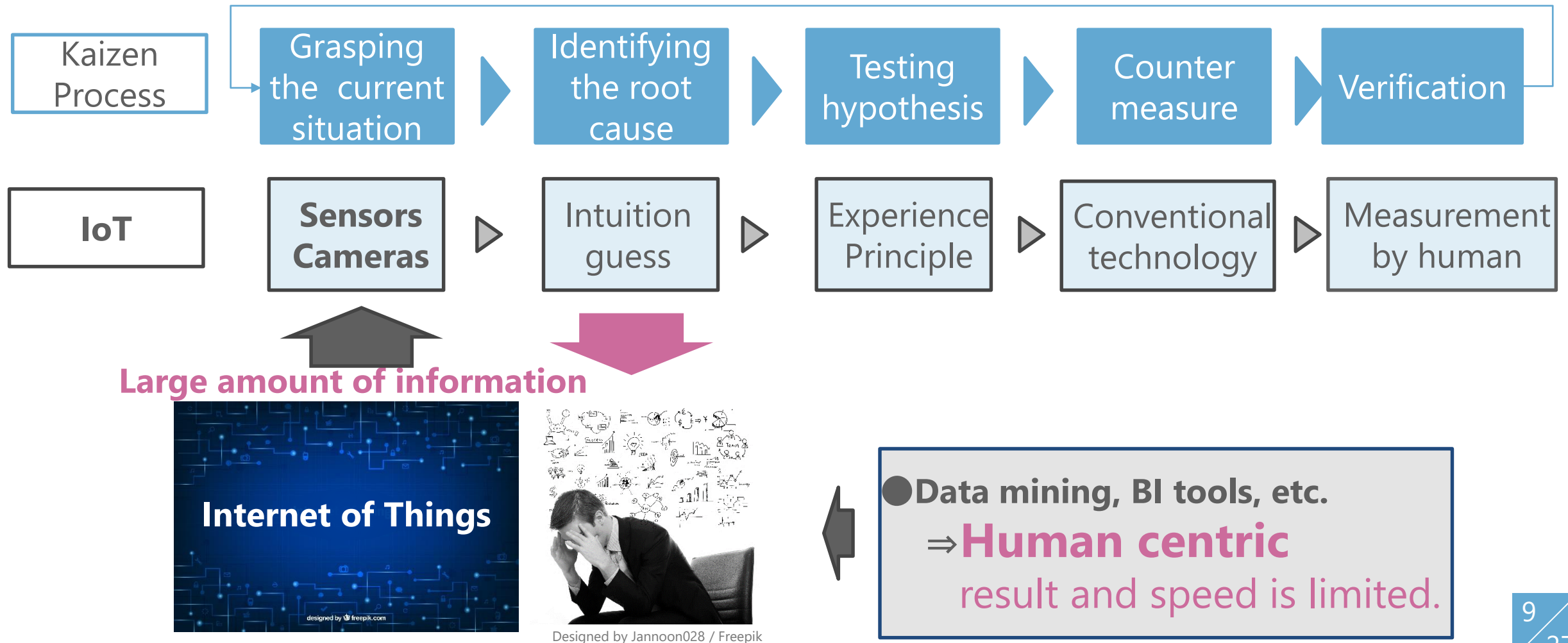
Kaizen (business improving) process so far  
engineer's insight, intuition, guess etc.  
⇒ Total process speed depends on humans processing speed.



## 2.2 “Monodukuri Innovation” (Level 1)

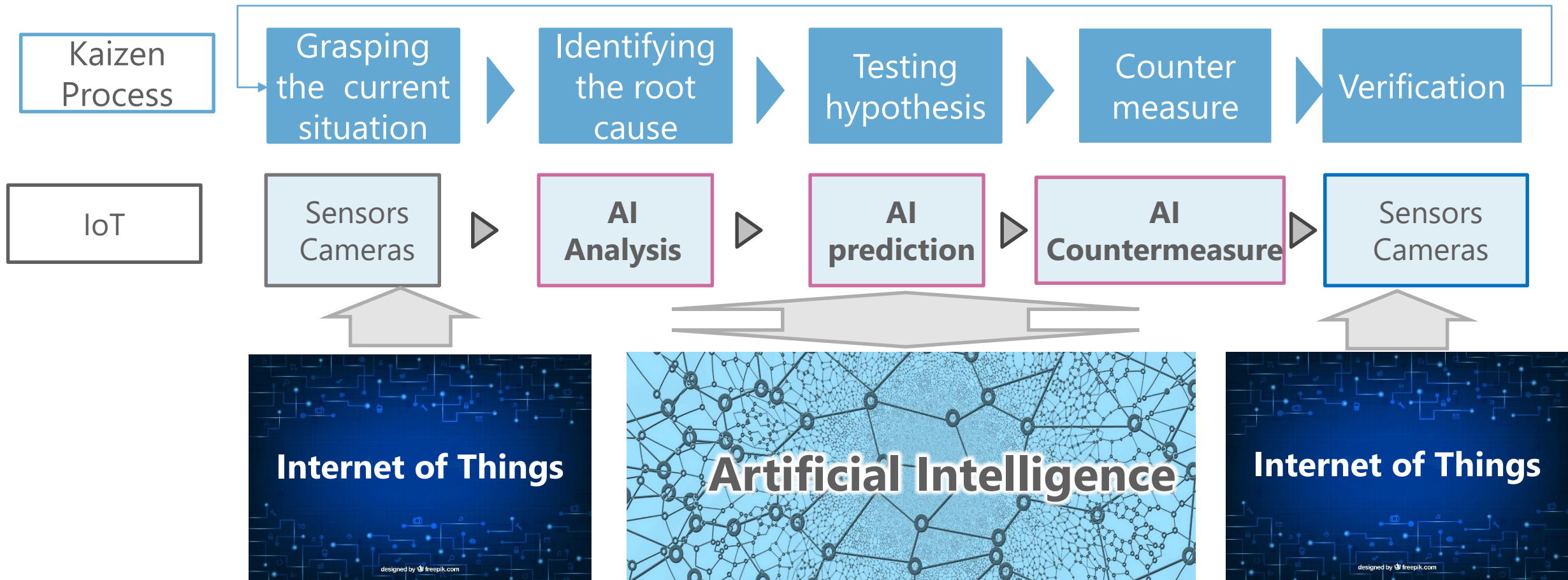
### ● Kaizen process with IoT (without AI)

Human centric process cannot deal a large amount of various sensors data



## 2.2 “Monodukuri Innovation” (Level 1)

- Unprecedented speed up, effectivity of countermeasure  
⇒ AI & IoT is complementary relationship. **Promote both as one entity.**



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# 3.1 Fiber Laser

- A fiber laser has excellent beam quality, high efficiency and high reliability.
- Laser diodes are key components of a fiber laser

## Fiber laser products



## Fiber laser Application



cutting



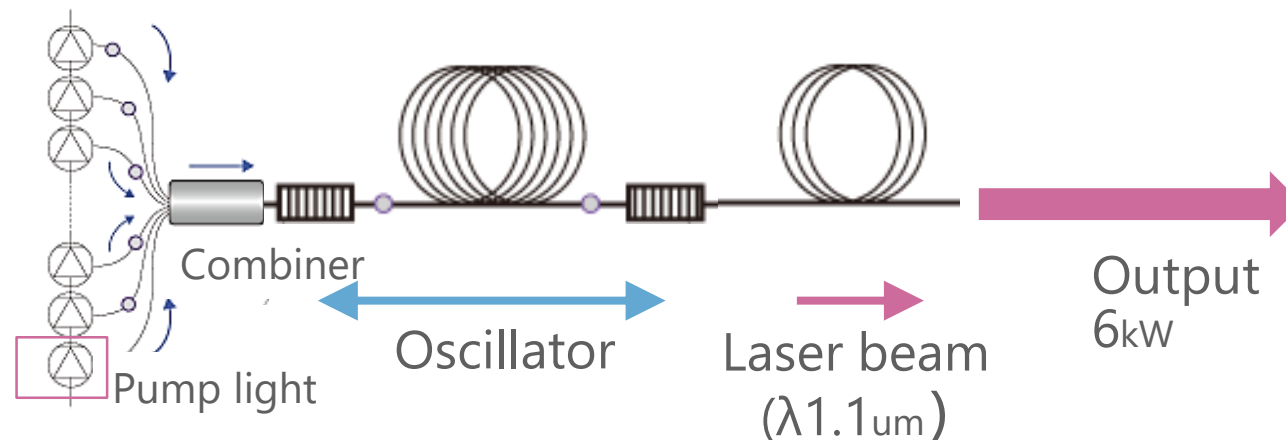
welding



Surface processing

### Fiber Laser Basic Configuration

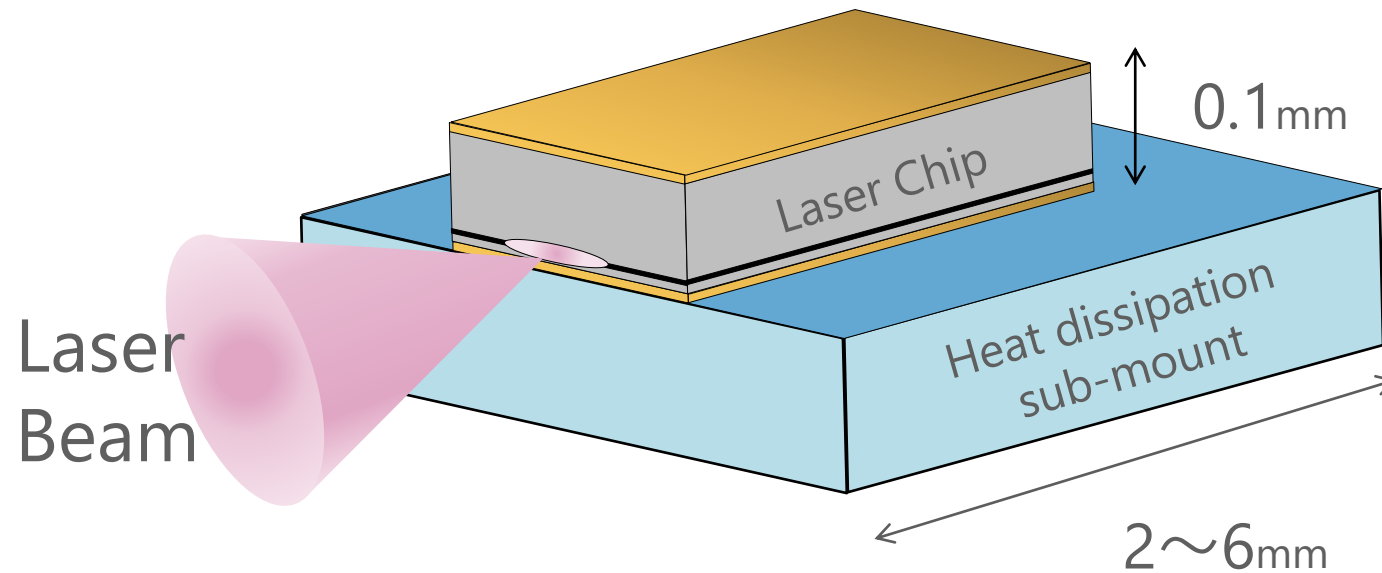
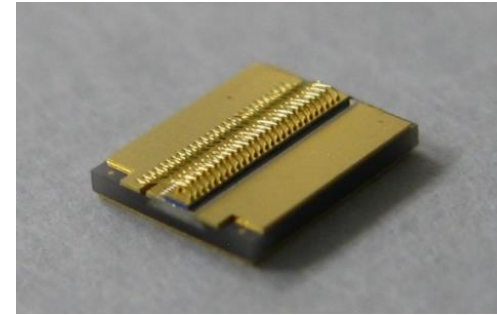
High-power semiconductor Laser



## 3.2 Pumping LD for Fiber Laser

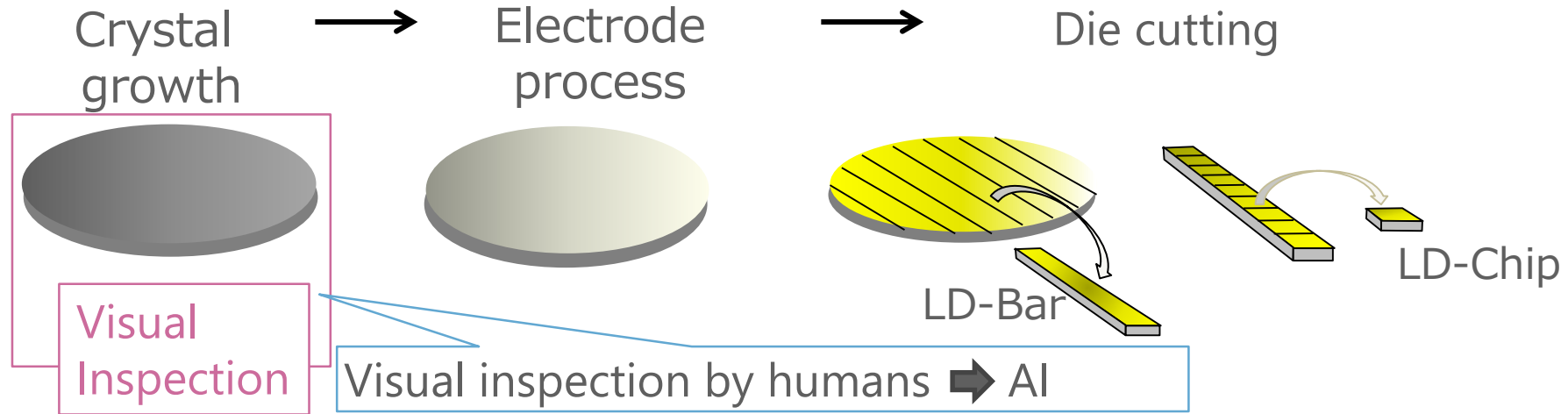
$\lambda: 900\text{nm}$

Output power: 10~20w

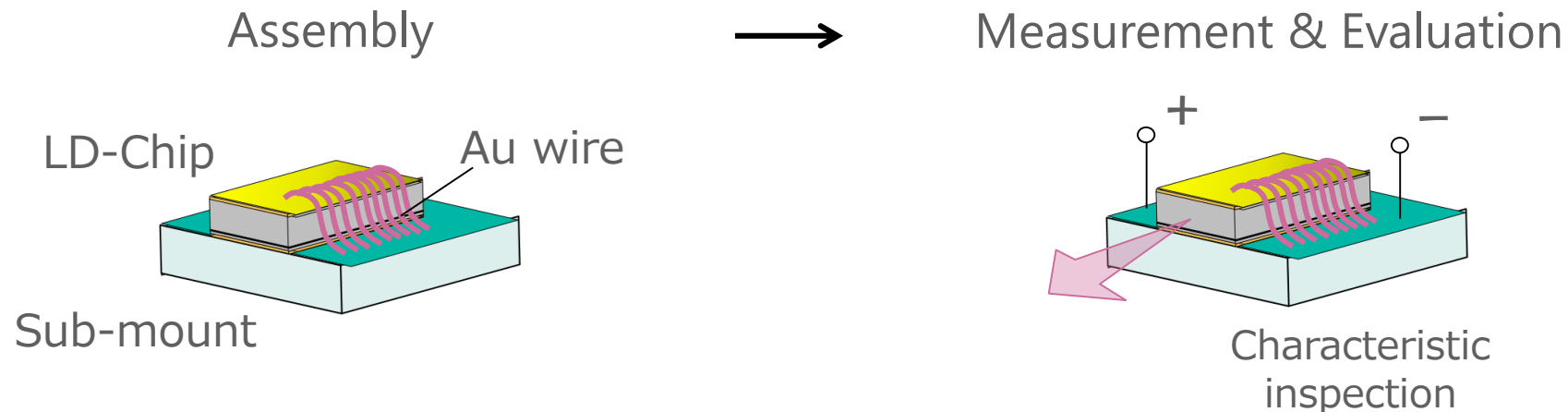


# 3.3 Manufacturing process of LD for fiber laser

## Pre process



## Post process





## 3.4 Visual inspection Criteria of LD chips

- 1.LD chips are classified into 5 categories (A-E) depending on their defects.
- 2."Others (defects) mode" needs to be treated and they are classified into different categories depending on their size and position.
- 3.For multiple defects in an LD chip, it is classified according to a priority of category classification.

		Defect modes			
		mode1	mode2	Others mode	
				Large	small
Occurrence position	Area 1	Category B	Category C	Category D	
	Area 2			Category E	
	Area 3	-	-	Category A	

"Others mode" category classification

-For multiple defects in one chip, categorized accordingly to the prioritized category.

Category C > Category D > Category E > Category B > Category A

A LD chip with no defect is categorized into A

This classification was done by skilled workers. ➡ AI



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# 4.1 Overview of visual inspection

## ●Purposes

- 1.Promotion of automatic inspection
2. Accumulation of knowledge and skills of AI

➡ **Improve competitiveness of manufacturing company (Monodukuri Innovation)**

- 3.Productivity and quality improvement of semiconductor lasers

➡ **Improve competitiveness of Fiber Laser products**

## ●Organization of Project Ai

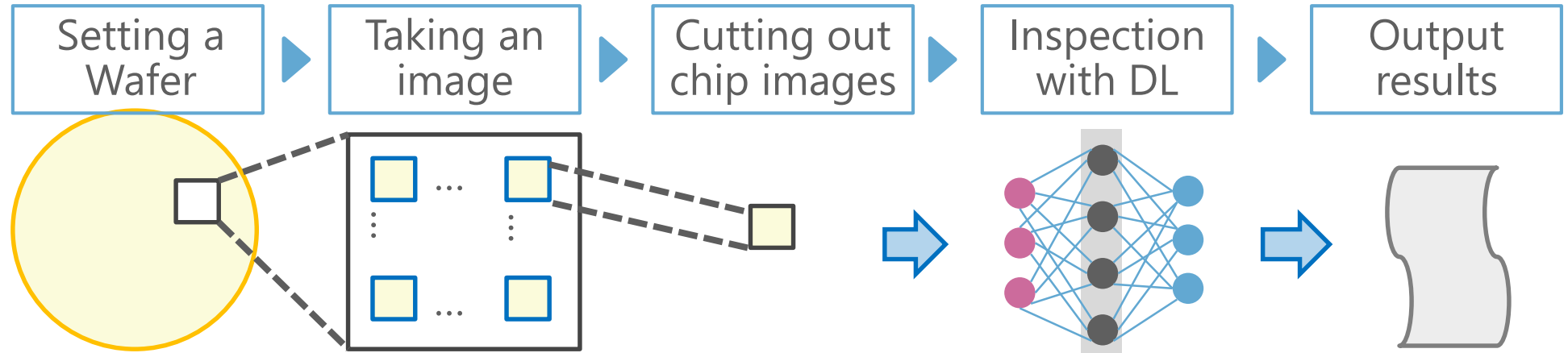
1. User (Image preparation)
2. AI technology development
3. System development

Optenergy, Inc. (Fujikura group)

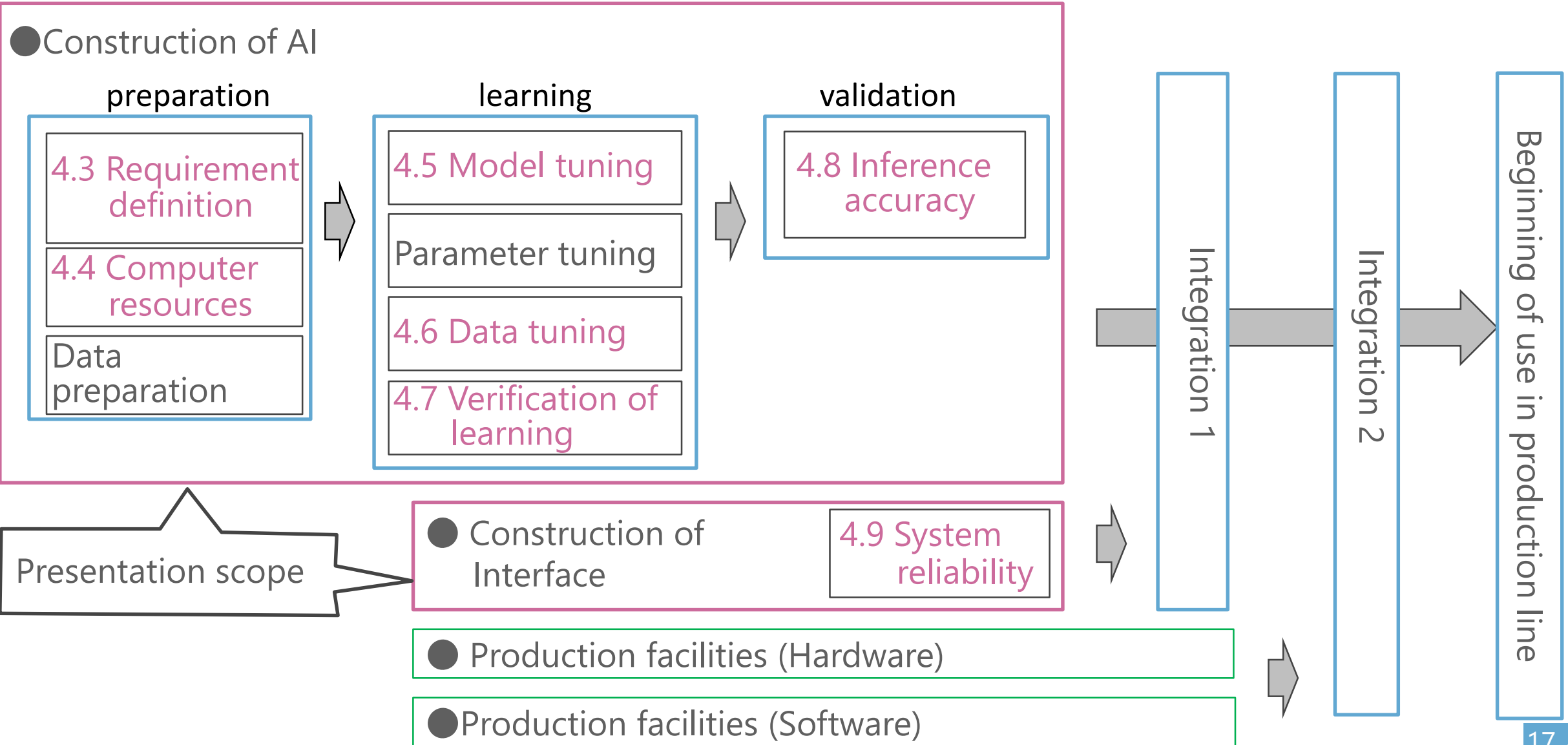
Fujikura Ltd. R&D dep.

Fujikura Ltd. Production facilities dep.

## ●process overview



## 4.2 System development process



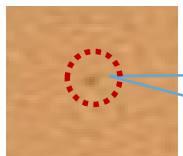
## 4.3 Requirement in visual inspection system

- The special requirement of our DL system different from ordinary DL is shown.

No	Special requirements	Action, issue	Slide
1	The large ratio of chip size to defects size( 1:500000) → Size is 30times larger	Pixel size Computer resources	4.4
		Deep learning model for large image	4.5
2	There is "Others" mode	"Others" modes is classified into sub modes.	4.6
3	Defects is classified into different class depending on the size and position	Create data base to manage image data	
4	There are few images in some failure categories.	data augmentation	
5	Explanation for AI classification	heat map	4.7
6	Implementation in production line	High reliability	4.9

## 4.4 Pixel size and computer resources

● Pixel size  $\Rightarrow$  minimum defects: 2x2 pixels, the whole chip: 2million pixels



Minimum defects 2x2 pixels  
 $\Rightarrow$  Minimum size



The whole chip

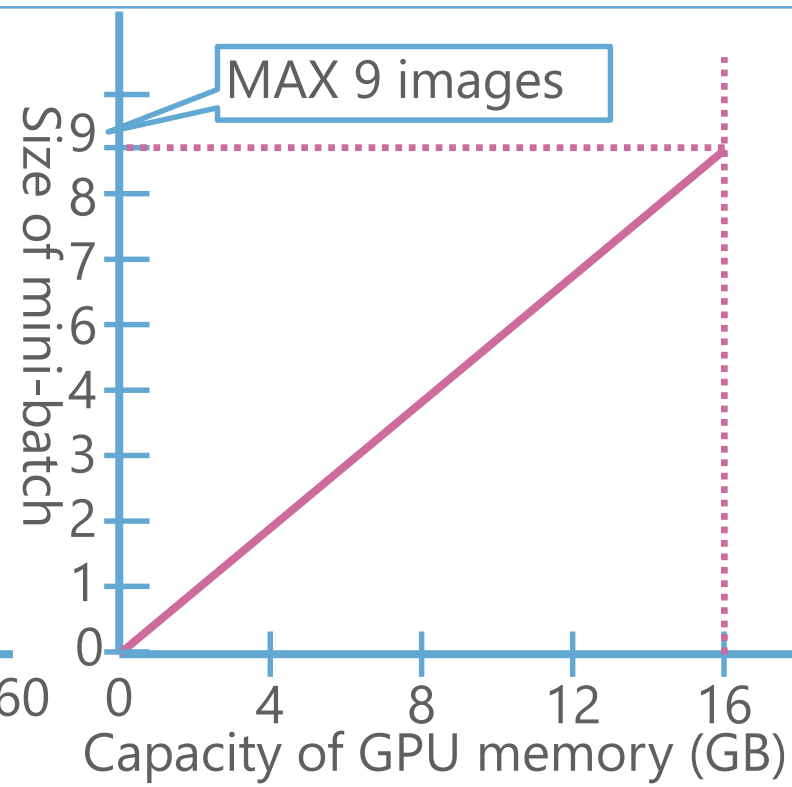
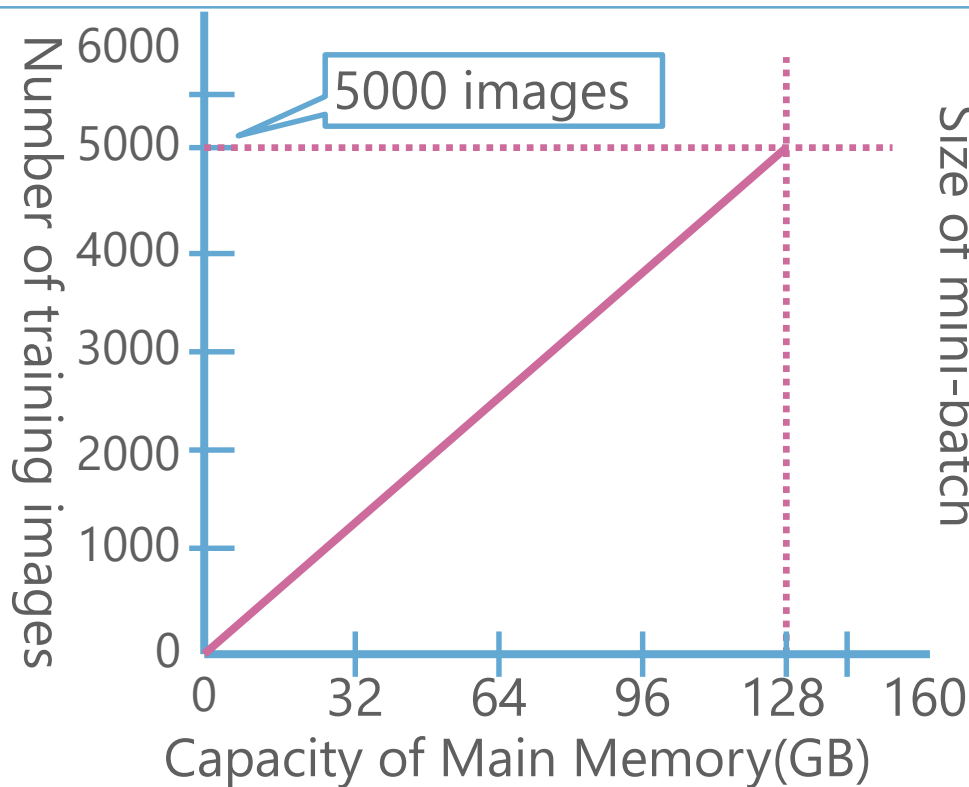
2million pixels  
 $\Rightarrow$  30 times large size  
(typical case:  $256 \times 256 = 65$  thousands)

Half a million  
times large



### DGX-1 with Tesla V100

Performance (GPU FP16)	1 peta FLOPS
System memory	512GB
GPU memory	16GB/GPU

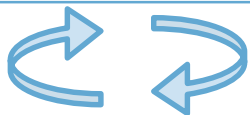


# 4.5 Tuning the network model

● General deep learning model doesn't work → Network model is build from scratch

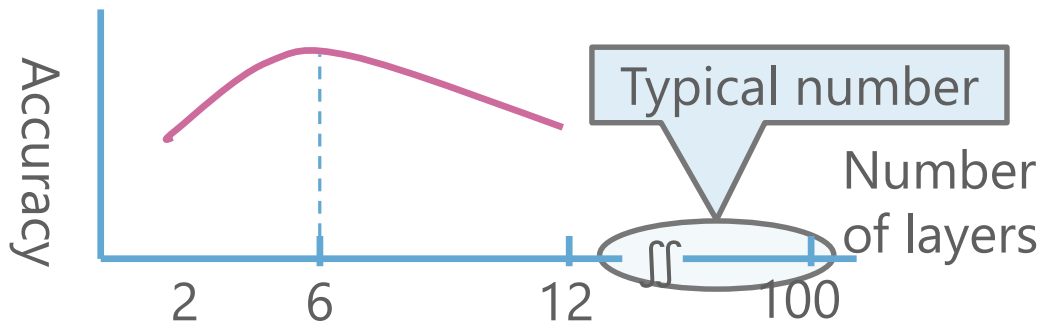
General model (ex. VGG) → NG

Varying the layer, filter size, etc.

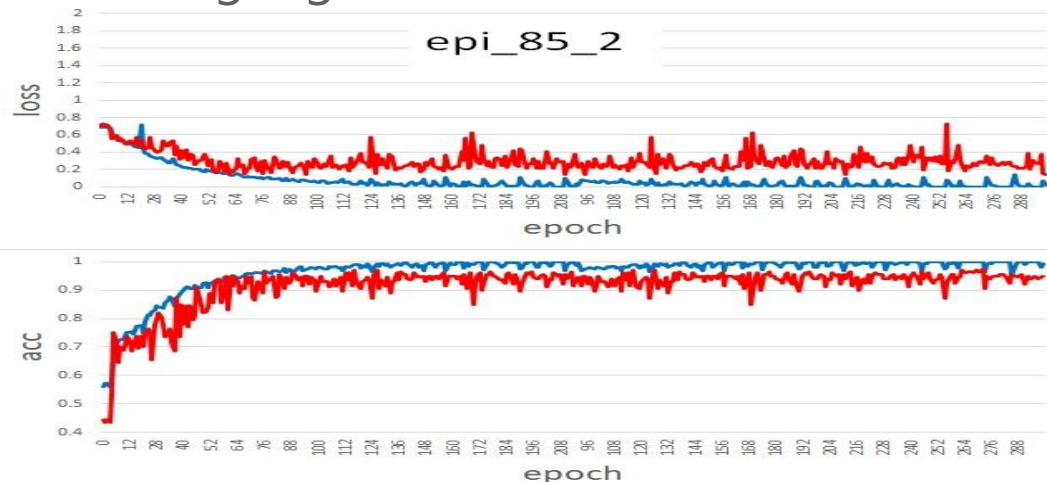


Checking Training log & heat map

Select the best network config.



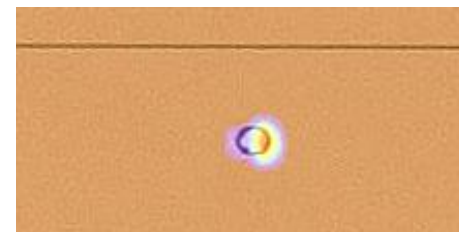
● Training log



Original image



with feature heat map

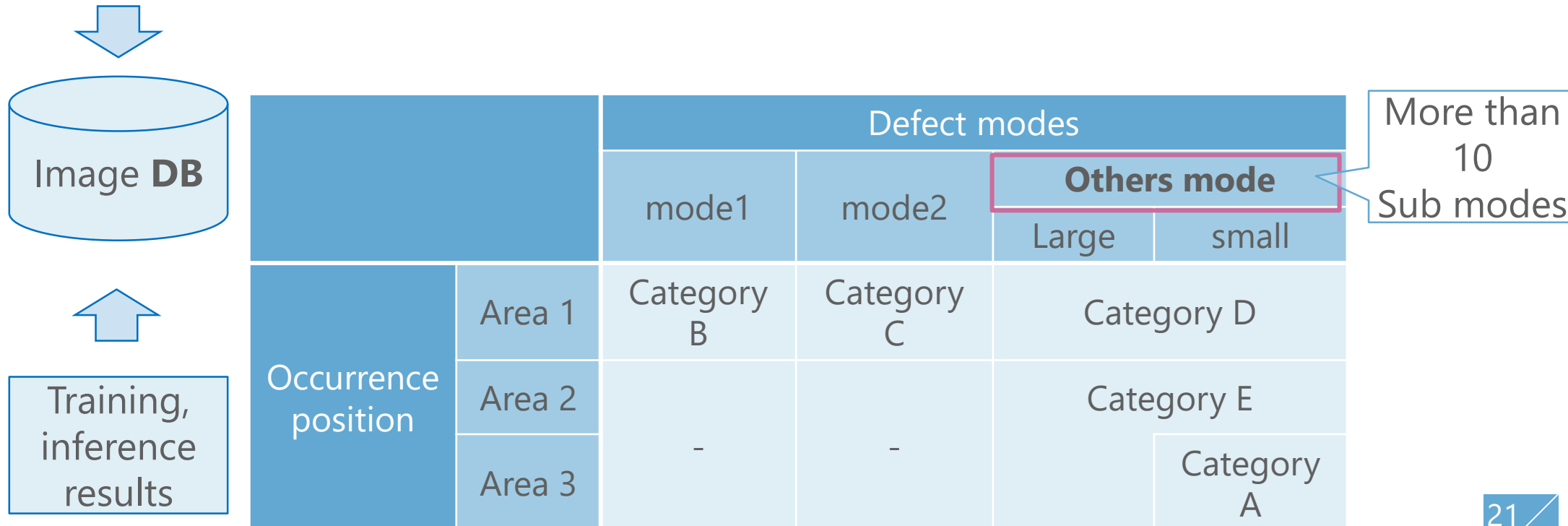


## 4.6 Data tuning

- Create **data base** to manage image data and utilized for improvement.
- For insufficient image data sets, data augmentation (LV1, 2) is performed.

- Data base item (example)

Chip ID	Lot No.	Defect mode	<b>Defect Sub mode</b>	Defect Position	Defect size	Augmentation	Inference results	...
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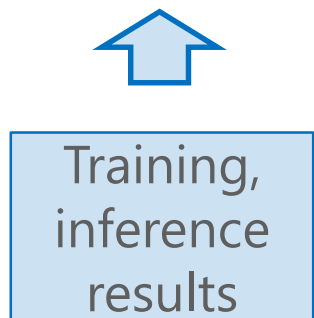
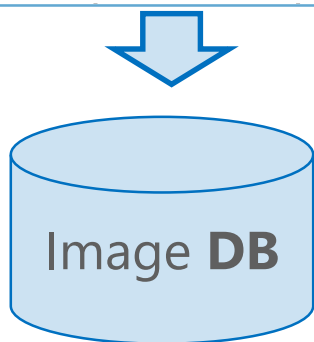


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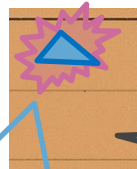
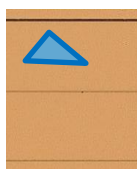
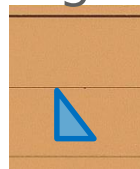
Chip ID	Lot No.	Defect mode	Defect Sub mode	Defect Position	Defect size	Augmentation	Inference results	...
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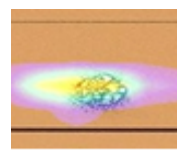
- LV1: General data augmentation (Rotation, up-down, Left-Right inversion, etc.)

- LV2: **synthesized image data is used.**

Original      Synthesized + heat map



Confirm **No adverse effect** due to synthesis images



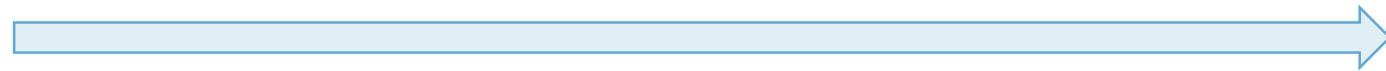


## 4.7 Verification of learning

- Not only the judgment of classification accuracy  
confirm that the heat map shows the correct position.
- Improve accuracy by eliminating discrepancies one by one

1.Human Judge  
=X

2.DL inference  
=X



Completion

# 4.7 Verification of learning

- We check the heat map for consistency during the inspection.
- Improve accuracy by eliminating discrepancies one by one

DL judge this cat. X by chance based on different point.

1. Human Judge  
= X

2. DL inference  
= X

3. Feature map

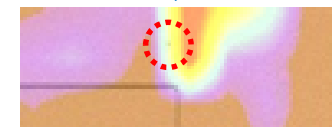
DL inference   human



discrepancies

DL focuses on a different point from the human's judge point.

4. Improvement  
Inferring what kind of error occurs from feature map  
⇒ Increase the similar images (including synthesized images)  
⇒ Confirm the effect

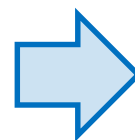


## 4.8 Inference results

- The 98% high accuracy has been achieved. That exceeds human accuracy (95%).  
⇒ Pass-Fail classification:98%, Category classification:95%

### ● Pass/Fail accuracy

Test data Pass:47% Fail:53%		DL system	
		Pass	Fail
Answer	Pass	46%	1%
	Fail	1%	52%



**Total:98%**

It exceeds Human classification accuracy (95%)

### ● Category classification accuracy

	Pass		Fail		
category	A	B	C	D	E
accuracy	97%	100%	95%	86%	88%



Consider actual occurrence distribution

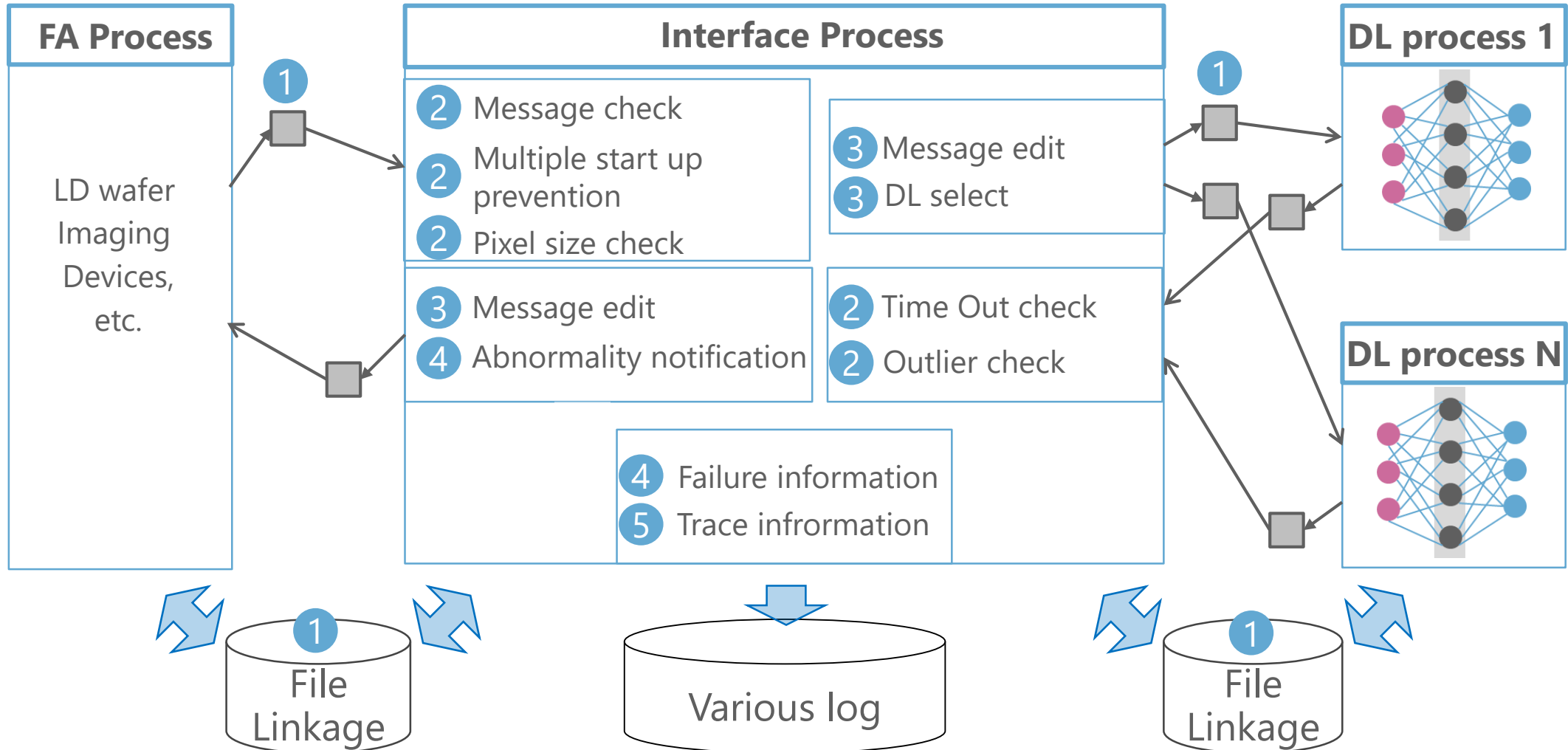
**Weighted accuracy  
95%**

# 4.9 System design for reliability, extensibility, etc.

● Implementation in production line



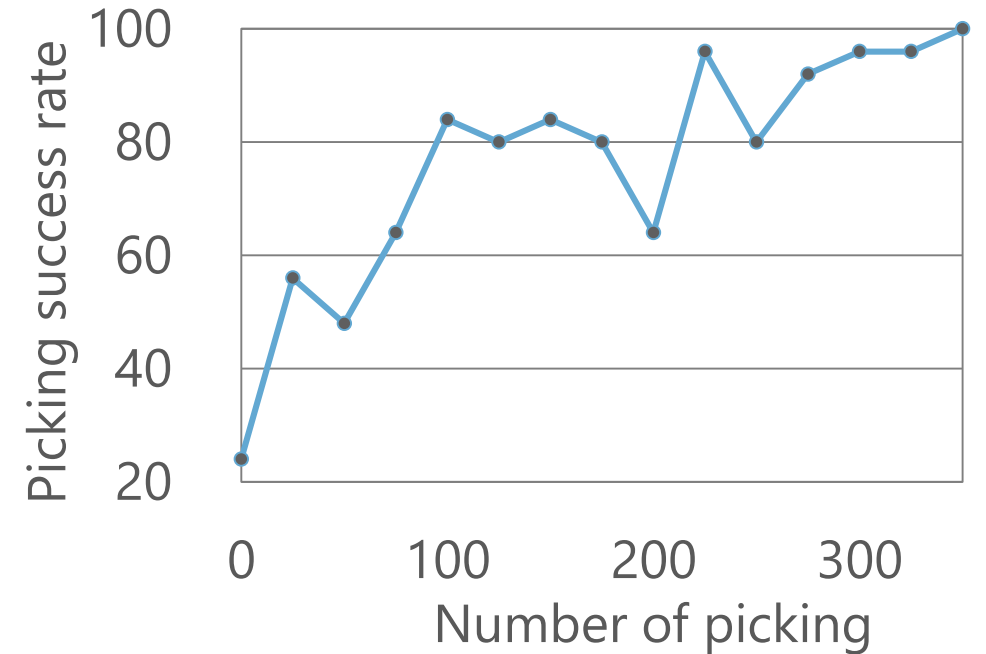
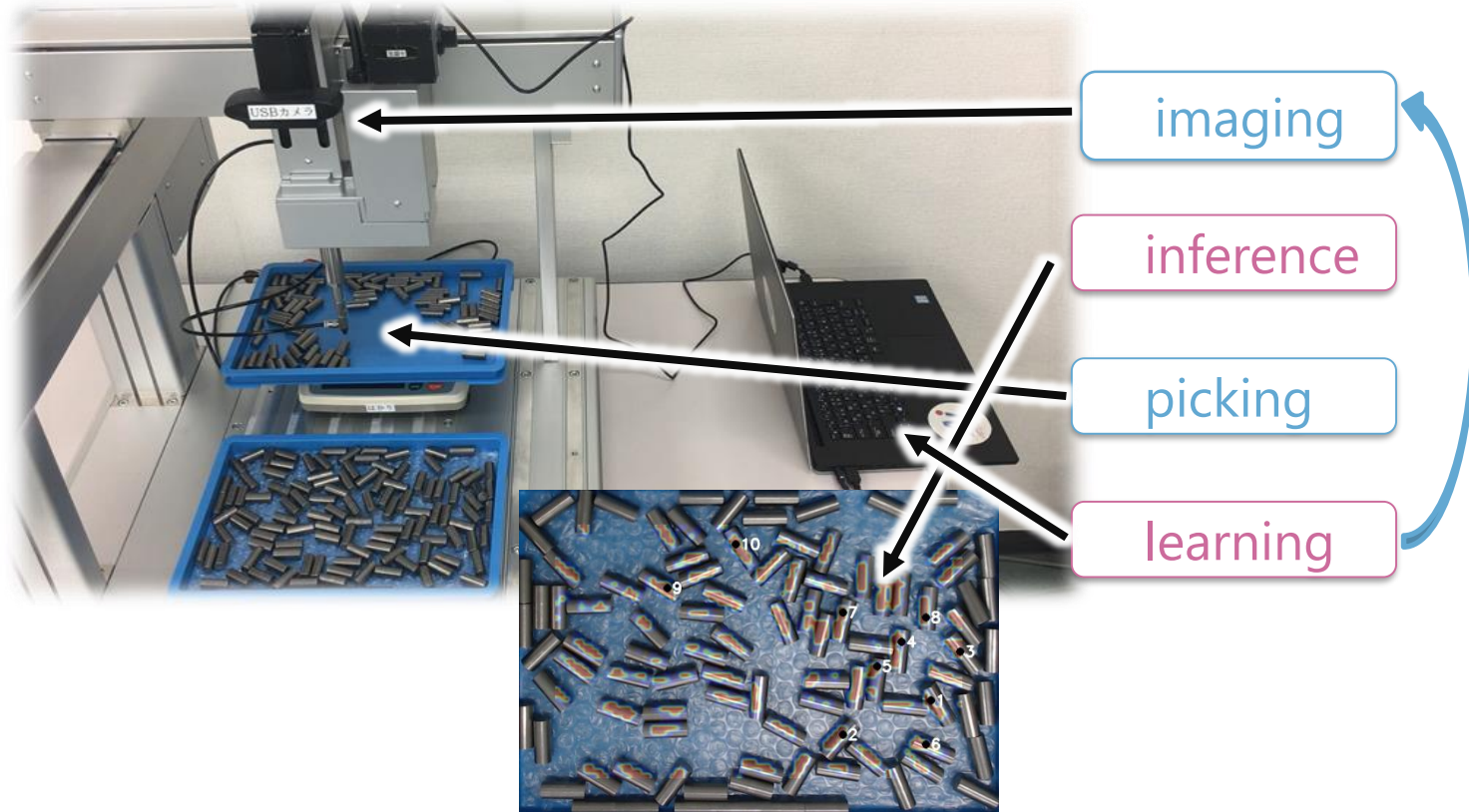
- 1 Loose coupling
- 2 High reliability
- 3 System extensibility
- 4 Failure recovery
- 5 Traceability



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# 5.1 Robot picking with AI

- The robot autonomously obtains the image of the objects, recognizes the image by deep learning, judges based on the recognition result, and decides the next action.



With 300 picking, more than 95% of success rate is achieved

- Future application of robot picking technology
  - ▪ ▪ Motion capturing, autonomous robot

## 5.2 Summary

- Fujikura is pushing forward “Monodukuri Innovation”
- Case study of AI
  - ① Visual inspection systems of LD with deep learning
    - Deep learning with large-scale images
      - The limitations of computer resources.
      - Model tuning for the original image dataset
    - Efforts to improve the accuracy
      - Managing the training image data by Database
      - Use of synthesized images for too little sub category.
      - Data tuning with feature heat map
  - ② Robot picking with AI
    - For future applications
      - Motion capturing, autonomous robot

# Thank you for your kind attention

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We are recruiting colleagues to work together with us!!  
[fjk.career@jp.fujikura.com](mailto:fjk.career@jp.fujikura.com)

