

# MACHINE LEARNING IN MANUFACTURING

Increase Quality and Reduce Waste  
with Automation



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# ABOUT ALLCLOUD



**3,000+**  
Successful  
Deployments



**6**  
AWS  
Competencies



**8**  
Global Operating  
Centers



**250+**  
Cloud  
Experts



**200+**  
AWS  
Certifications

Tel Aviv

Bucharest

Munich

Berlin

Vienna

New York

San Francisco

Toronto



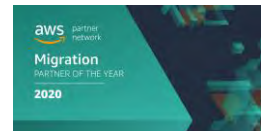
ISO 27001:2013  
27017:2015 Certified



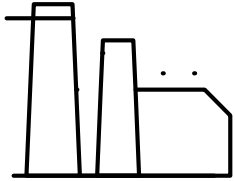
EU GDPR  
COMPLIANT



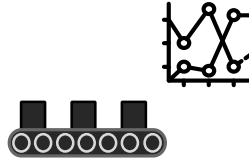
aws **200** certified  
AWS PARTNER NETWORK



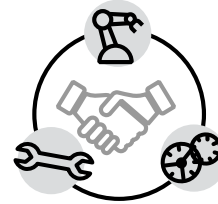
# MANUFACTURING USE CASES



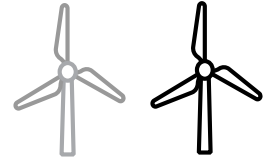
**Production/Process optimization**



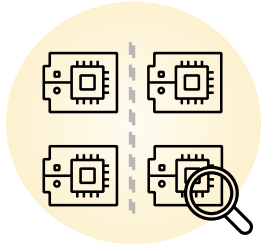
**Predictive maintenance for machines**



**Machine as a service**



**Digital Twin**



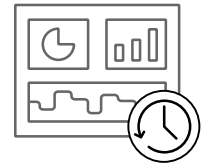
**Computer vision for quality control**



**Automated material management**



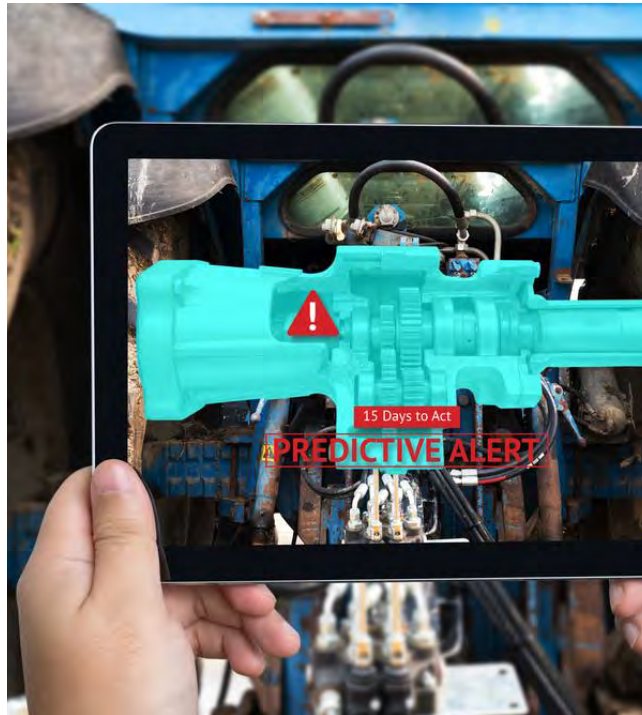
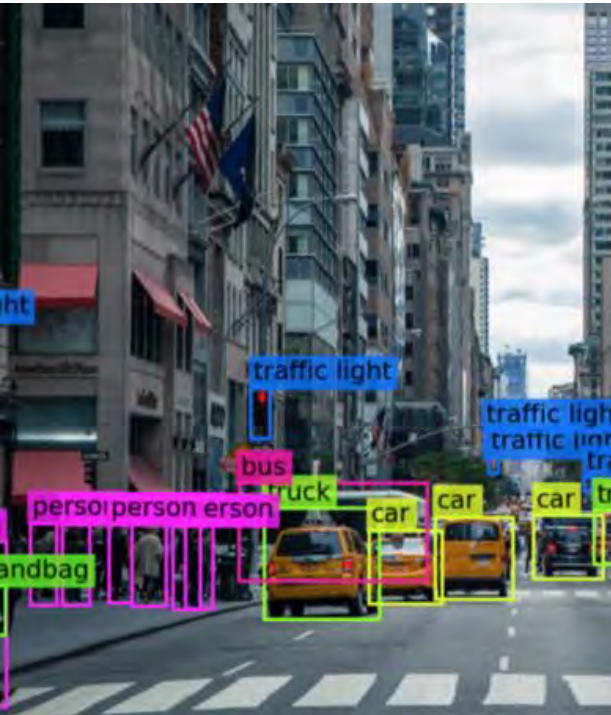
**Track & Trace**



**Data visualization**

# POSSIBLE APPROACH:

## Computer Vision for Quality Control (QC)





# USE CASE:

## Automated QC of Textile/Fabric



### Challenge

- Early detection of flaws in weaving.



### Goals

- Reduce manual QC efforts.
- Reduce number of flawed prods. from leaving factory.

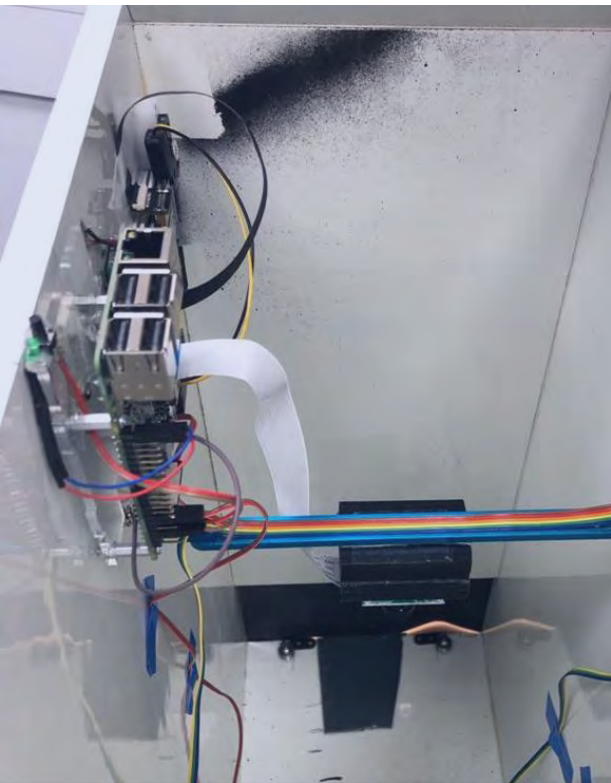


### Approach

- Self-developed IoT hardware (Raspberry Pi).
- ML using Amazon SageMaker.



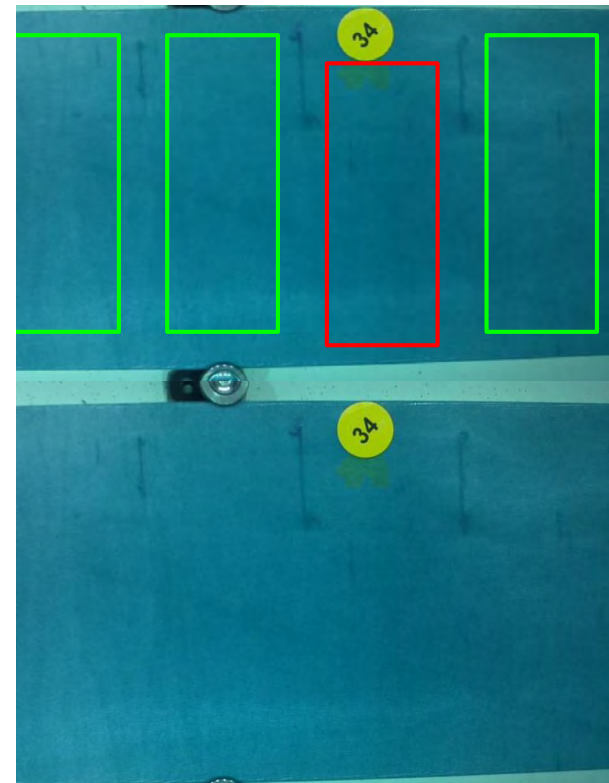
# QUALITY CONTROL



**PROTOTYPE**



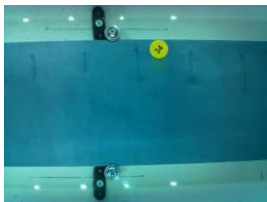
**DATA COLLECTION**



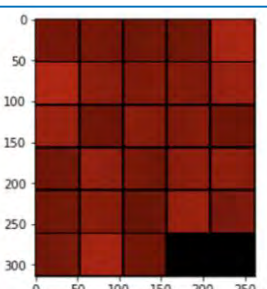
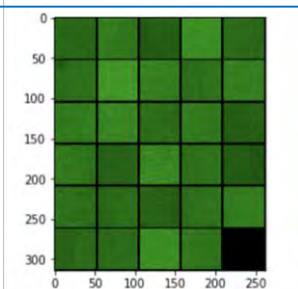
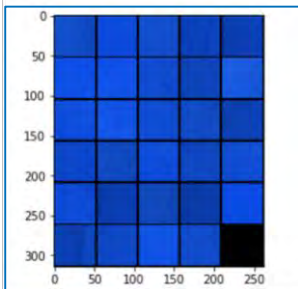
**LABELLING**

# COMPUTER VISION

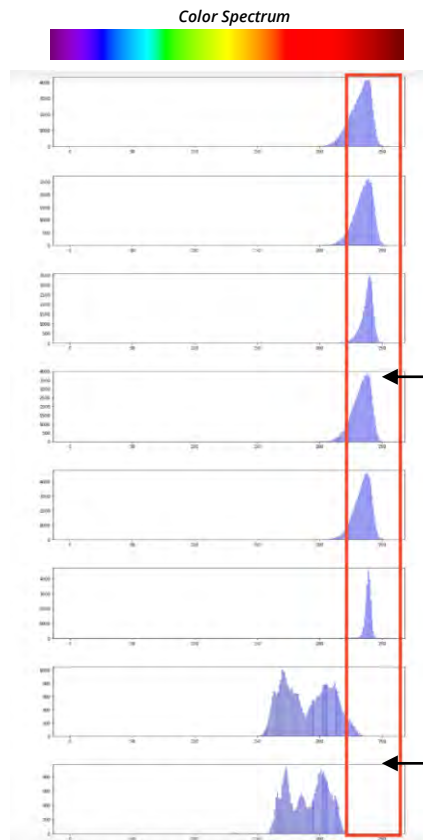
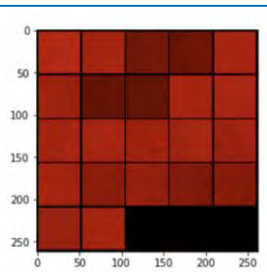
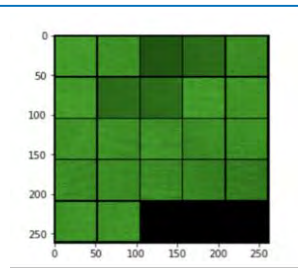
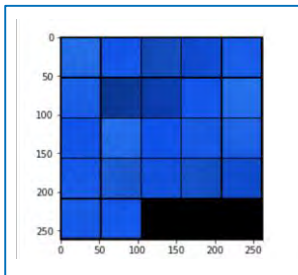
Which Approach / Model?



Baselines



Defects

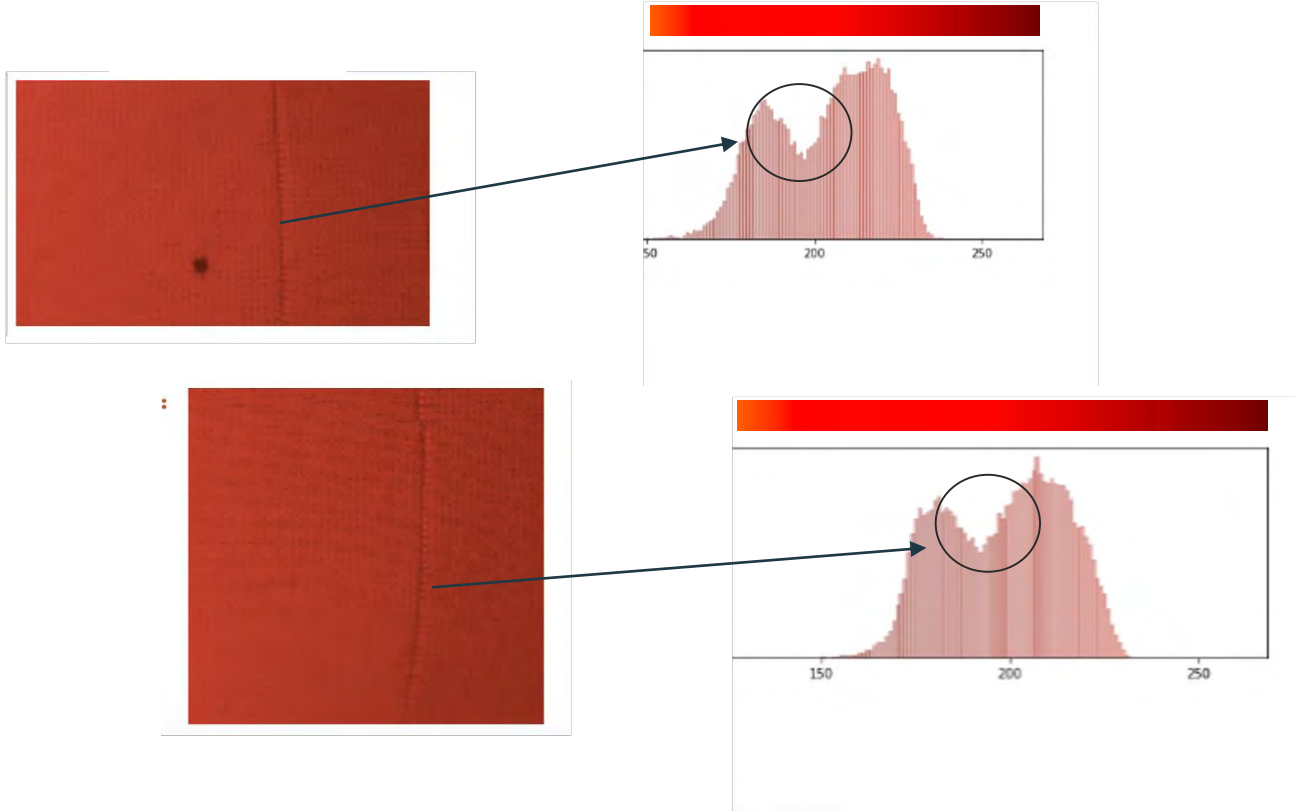


Baselines

Defects

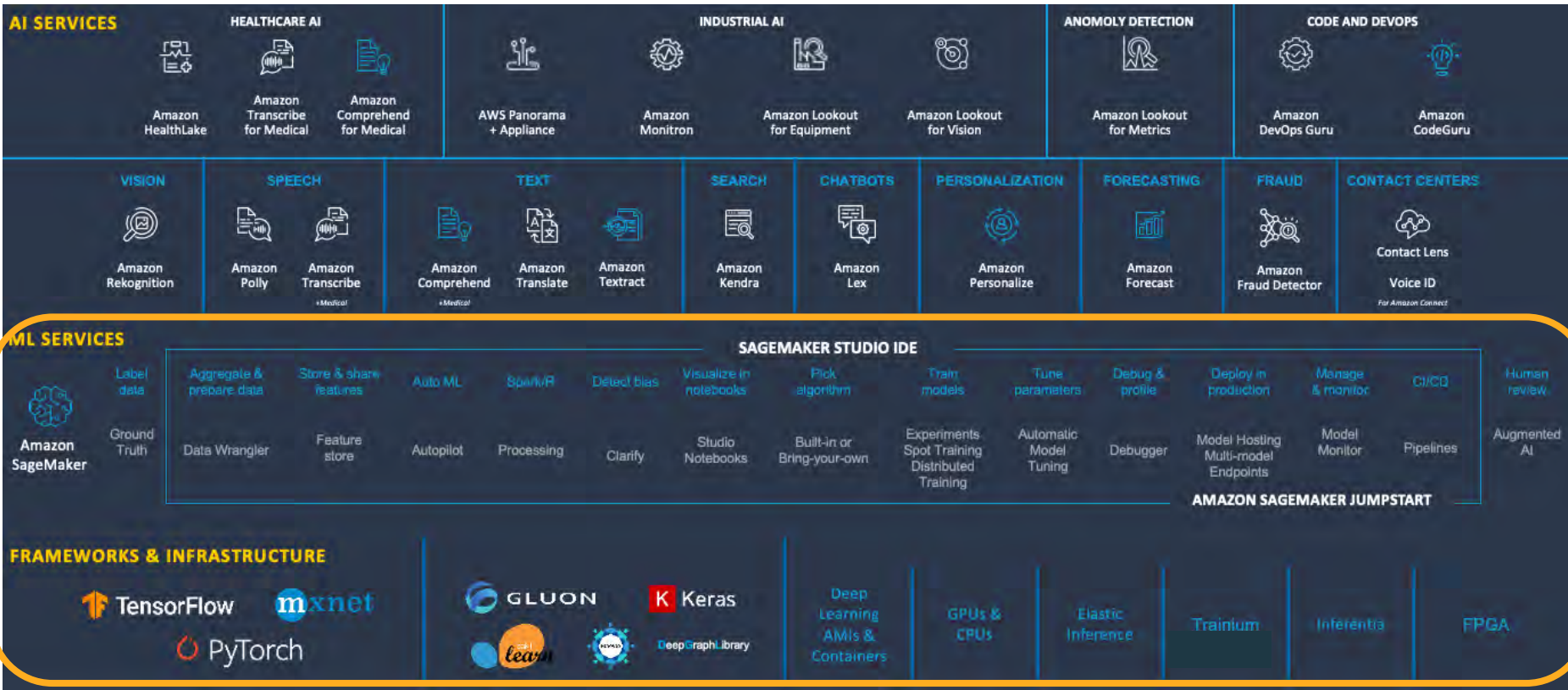
# COMPUTER VISION

## Distribution of RED for DEFECTS



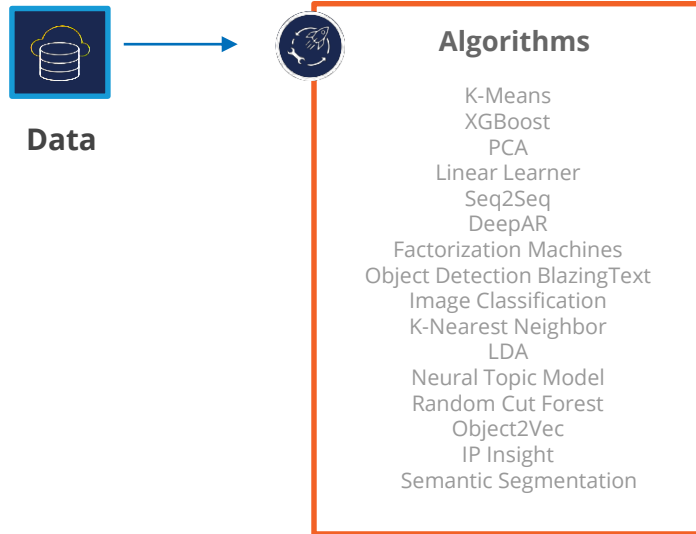


# ML/AI on AWS

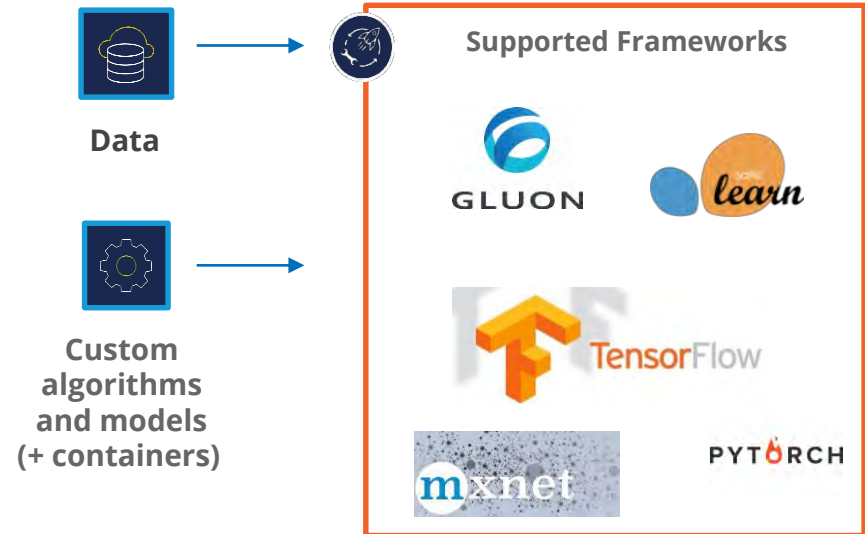


# AMAZON SAGEMAKER

## Built-In Algorithms

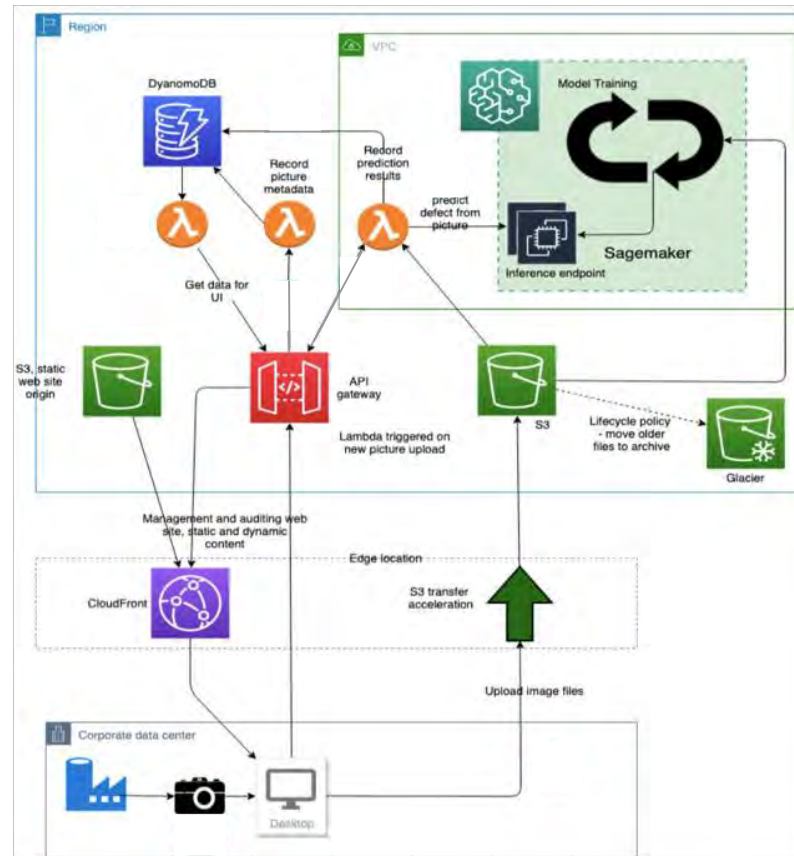


## Build/Bring-Your-Own-Code



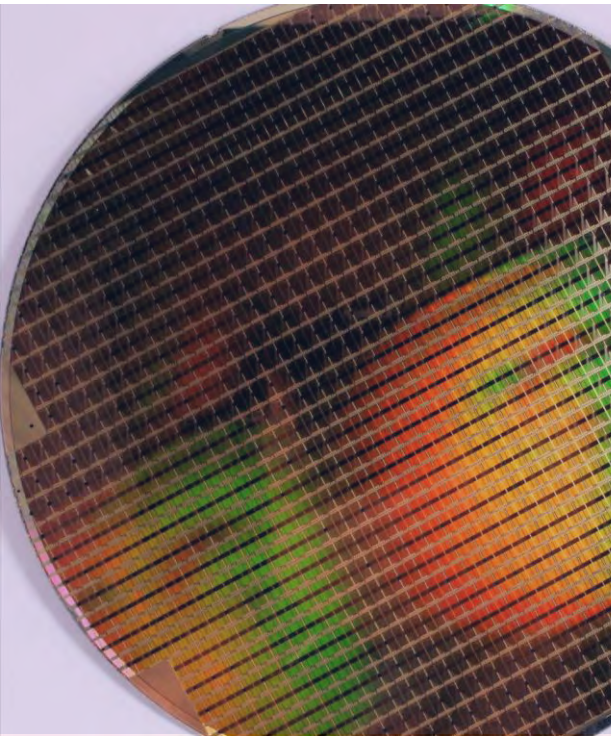
# FROM R&D TO PRODUCTION

- R&D, iterative, non-trivial, more open questions than in other IT projects:
  - Does it even work?
  - Visual data enough?
  - Which algorithms to use?
  - ...
- Amazon Web Services (AWS):
  - Inherently scalable ecosystem.
  - Fast, goal-oriented, perfect for tests and trials.
  - High-level building blocks.

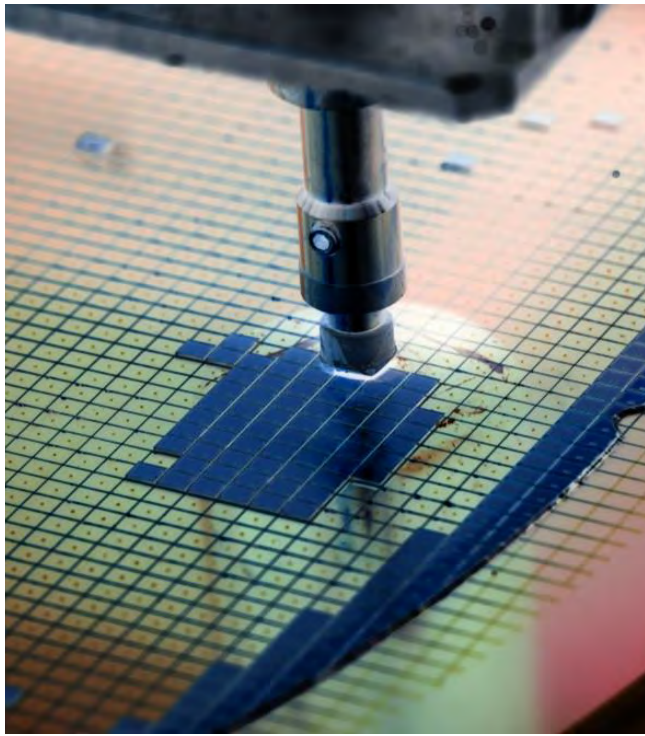


# USE CASE:

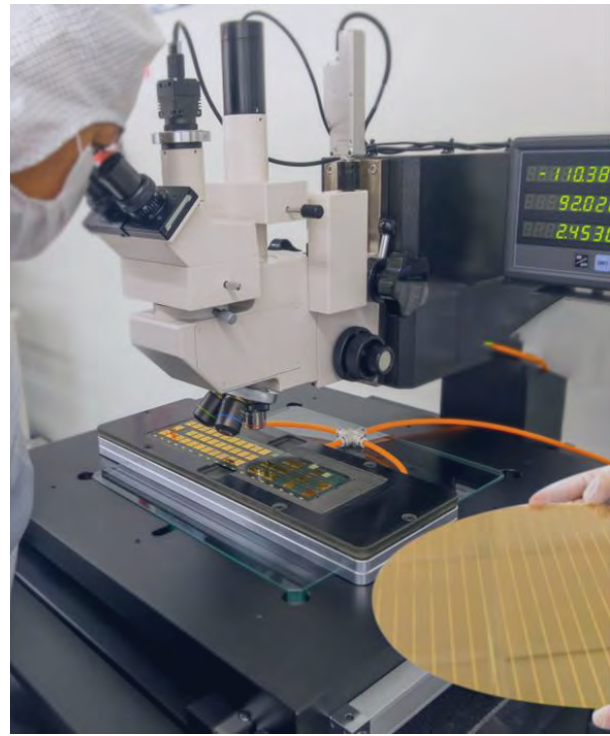
## WAFER QC



**SILICON WAFERS**



**CHIP PRODUCTION**



**DEFECTS DETECTION**



# USE CASE:

## Tower Semiconductor



### Challenge

- Automate QC from hi-res pictures.



### Goals

- Reduce downstream "surprises".
- Improve upon the existing system.



### Approach

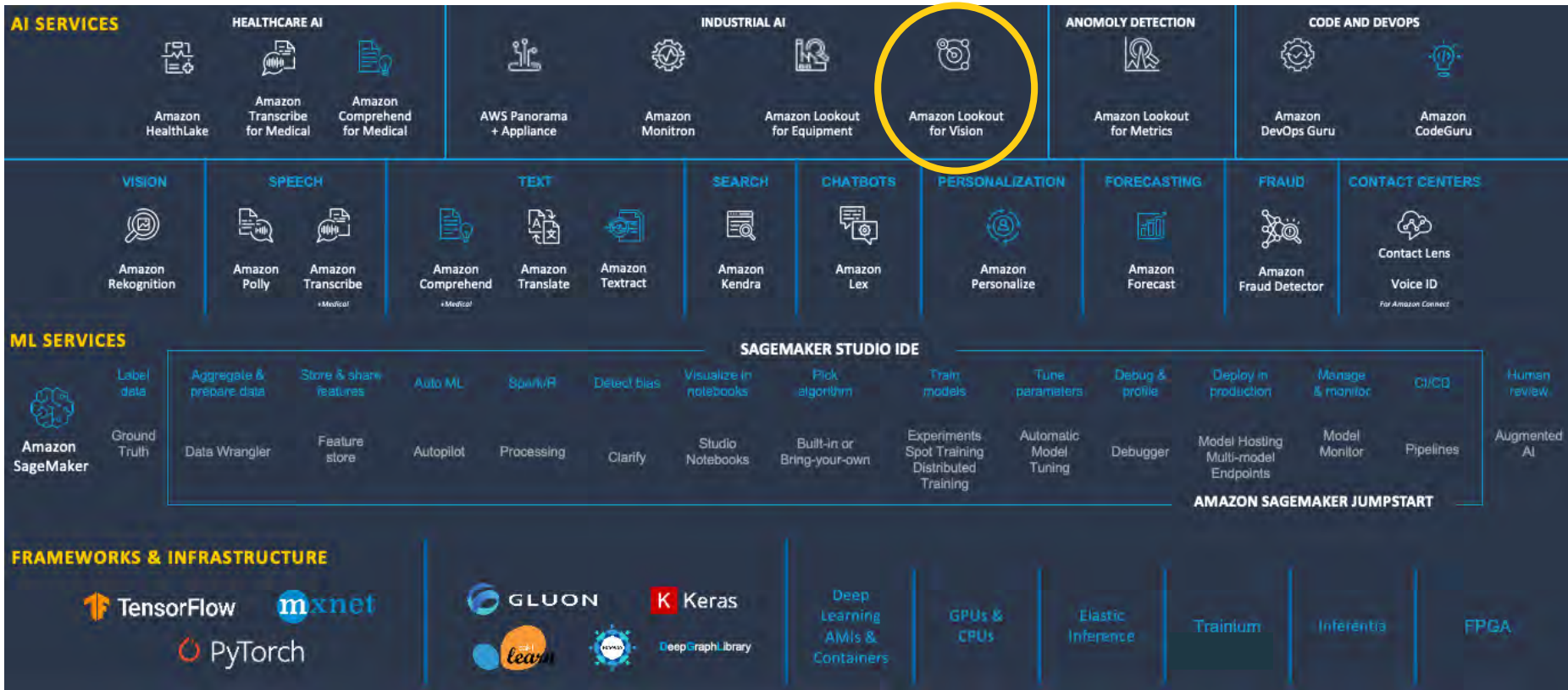
- ML approach based on SageMaker.
- Low / No code ML: Lookout for Vision.

**Tower**  
Semiconductor

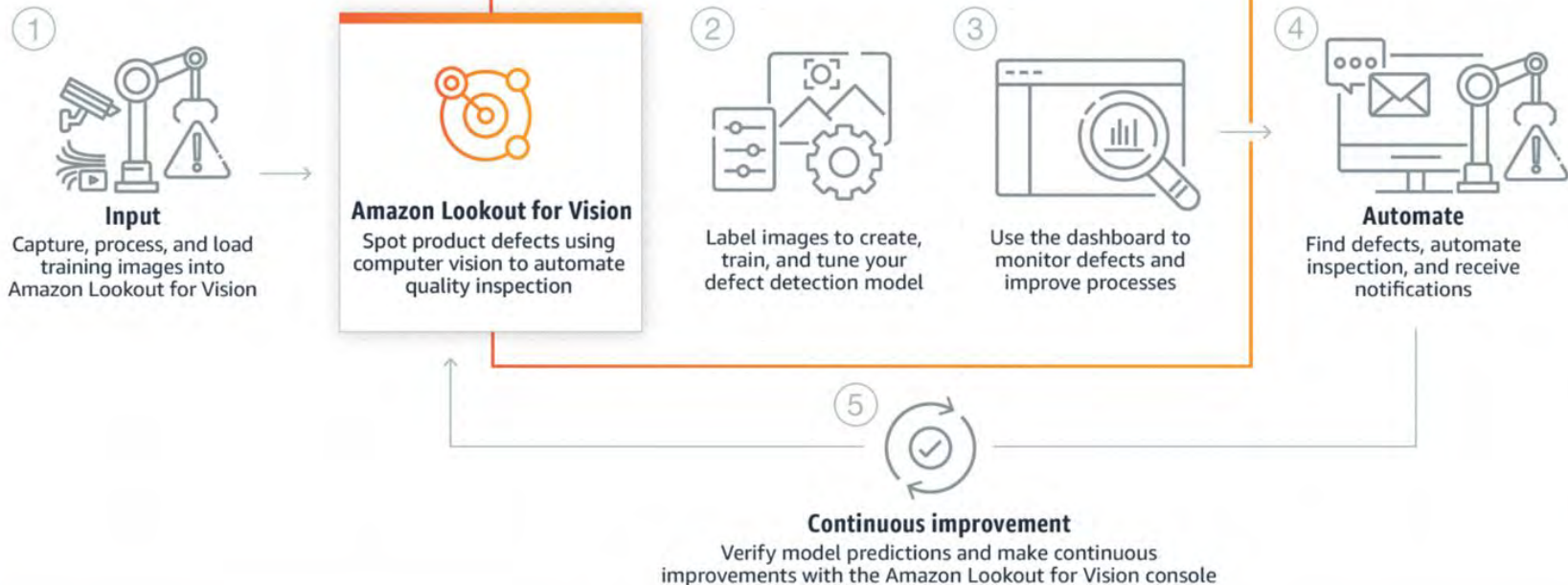
Where **Analog** and **Value** Meet



# ML/AI on AWS

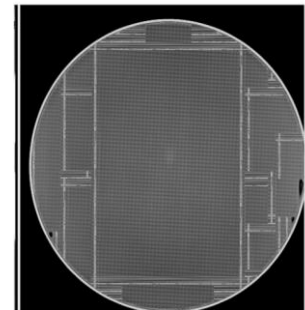
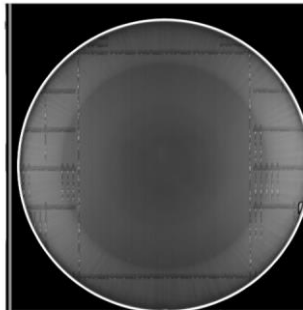
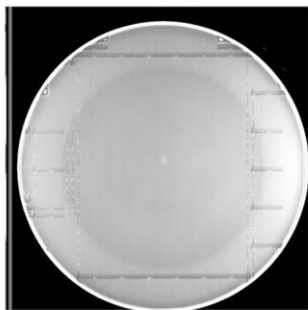
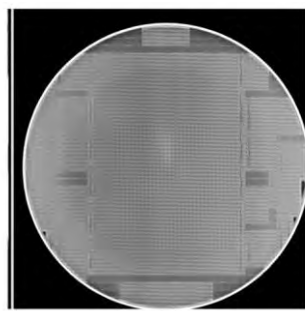
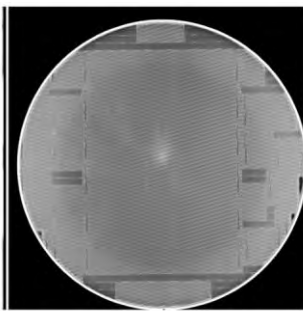
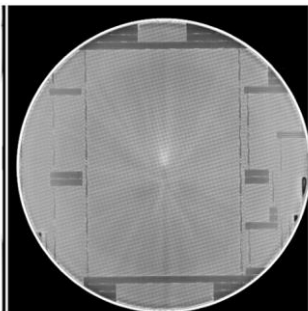


# AMAZON LOOKOUT FOR VISION



# USE CASE:

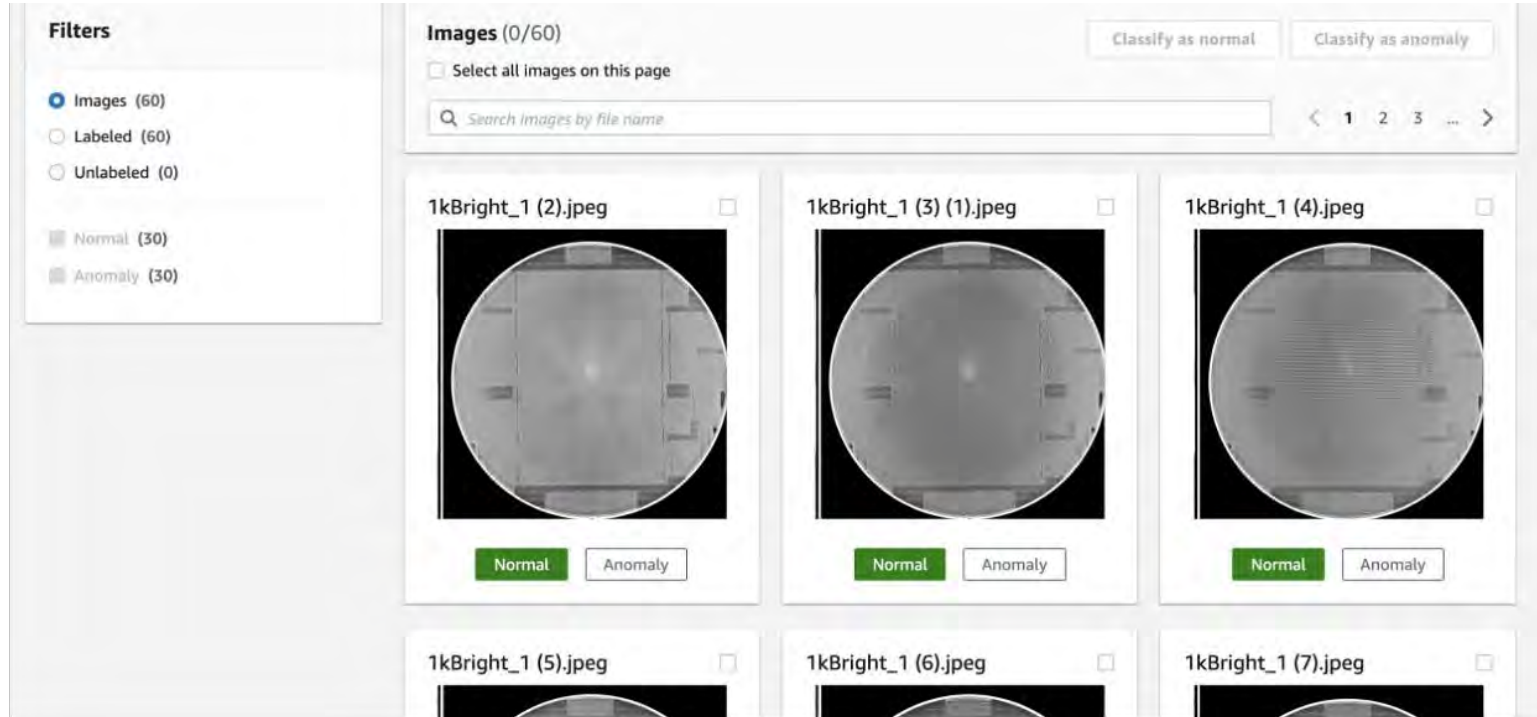
Tower Semiconductor





# LOOKOUT - LOW / NO-CODE

Labeling of known normal/abnormal wafers



# LOOKOUT - LOW / NO-CODE

## Training the model(s)

The screenshot shows the Lookout for Images training interface. At the top right, there are three buttons: "Add labels", "Actions", and "Train model". The "Train model" button is highlighted with a red rectangle. Below these buttons is a light blue informational banner with an information icon and the text: "You have enough labeled images to train a model." followed by two bullet points: "• You can improve the quality of your model by adding more labeled images." and "• Unlabeled images aren't used for training." On the left side, there is a "Filters" panel with the following options: "Images (60)" (selected with a blue dot), "Labeled (60)" (radio button), "Unlabeled (0)" (radio button), "Normal (30)" (checkbox), and "Anomaly (30)" (checkbox). The main area is titled "Images (60)" and contains a search bar with the placeholder text "Search images by file name". Below the search bar, there are three image thumbnails, each with a filename above it: "1kBright\_1 (2).jpeg", "1kBright\_1 (3) (1).jpeg", and "1kBright\_1 (4).jpeg". Each thumbnail shows a grayscale image of a circular object, likely a lens or a camera view.

**Train model**

**Filters**

- ☒ Images (60)
- ☐ Labeled (60)
- ☐ Unlabeled (0)
- ☐ Normal (30)
- ☐ Anomaly (30)

**Images (60)**

Search images by file name

1kBright\_1 (2).jpeg

1kBRight\_1 (3) (1).jpeg

1kBRight\_1 (4).jpeg

# LOOKOUT - LOW / NO-CODE

## Test performance of model(s)

### Model performance metrics [Info](#)

Status

🟢 Training complete

Status message

The model is ready for hosting

Date created

January 26th, 2021 at 12:32:43 PM

Train duration

54 minutes 57 seconds

Test images

24 images

Precision



13 anomalies were correct out of  
14 total predictions

Recall



13 anomalies were predicted out of  
13 total anomalies

F1 score



The overall model performance.

### Test results overview [Info](#)

#### Predicted as normal

Correct (True negative)

10

Incorrect (False negative)

0

#### Predicted as anomaly

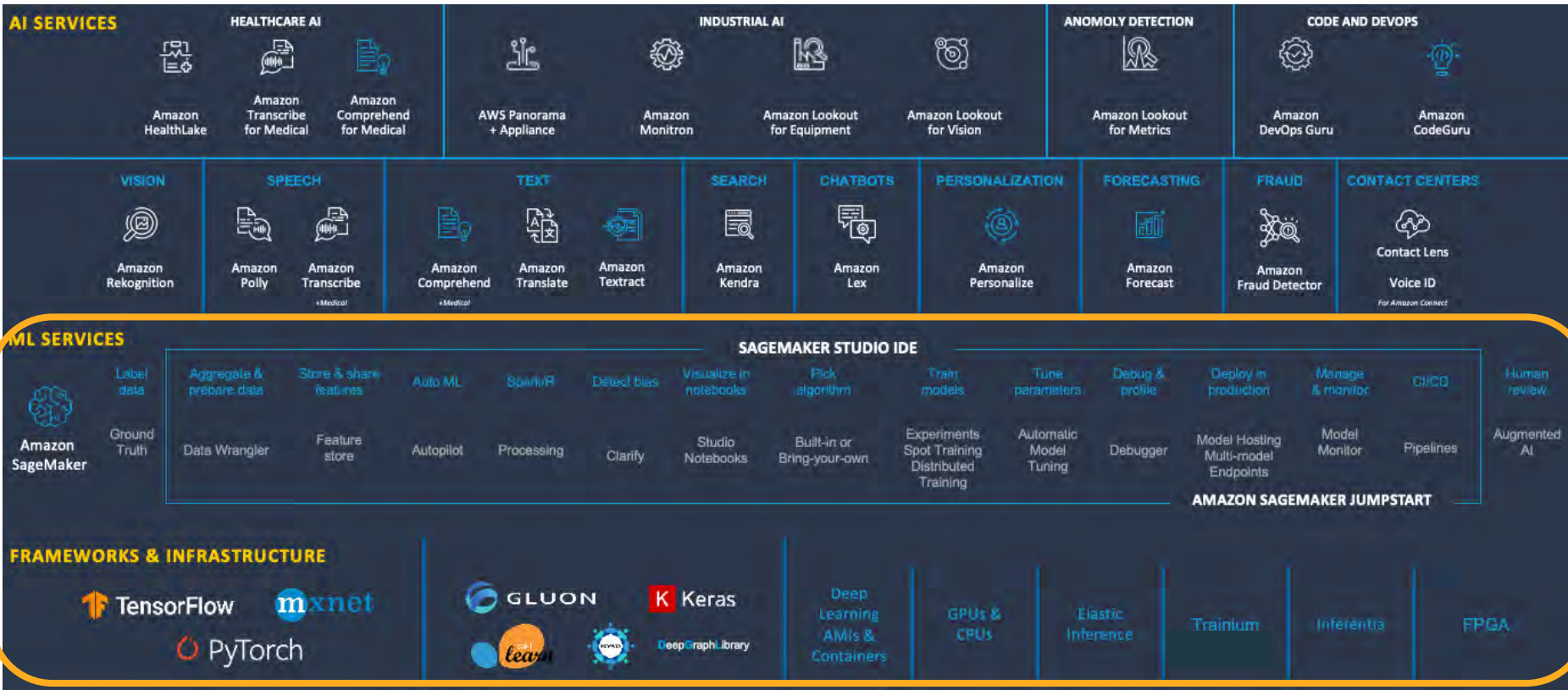
Correct (True positive)

13

Incorrect (False positive)

1

# ML/AI on AWS

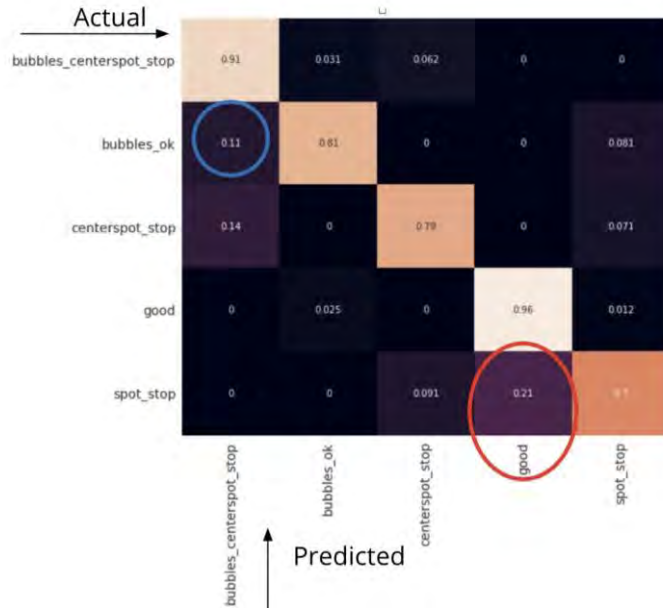




# USE CASE:

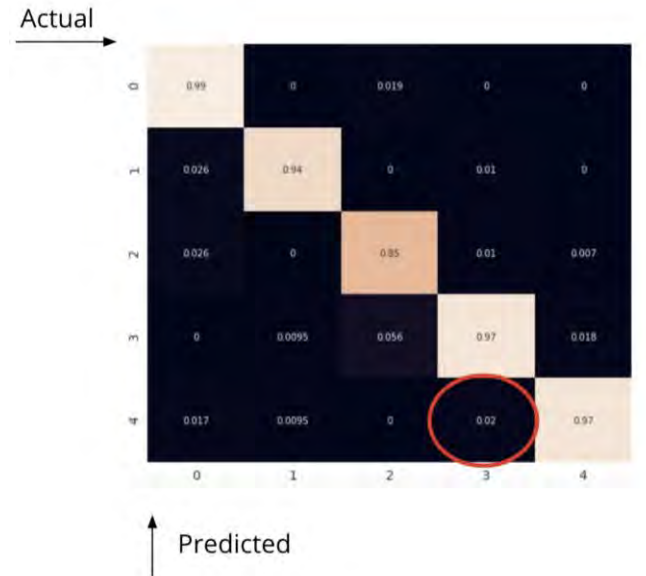
## Tower Semiconductor - Using Deep Learning / ANNs

**BEFORE**



Existing system:  
The error rate is **21%**

**AFTER**



With Machine Learning:  
Error rate reduced to **2%**

# AllCloud - DATA, ML / AI



## Data Pipelines (Lakes / WH)

Setting-up/Implementing data lake projects and related data services



## Data Analytics & BI

Setting up BI Solutions (or optimizing existing ones) on top of DWH/Lakes.



## Data Access Frameworks

Applications Integration, Data Products, Digital Access, APIs



## ML / AI - DS

SageMaker implementation, model & algorithms setup. Business cases: NLP Prediction, Computer-Vision



## IoT Solutions

Connecting devices, Streaming data, Implementing end-point AI



## Data Operations

Managed-Service for Data Operation Center - Ensuring system productivity

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>> WE ARE HIRING <<



AllCloud