

# Automated Visual Inspection (AI and Machine Learning)

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Process Development Scientific Director



**INSPIRING  
INNOVATION**

Transformation & Future

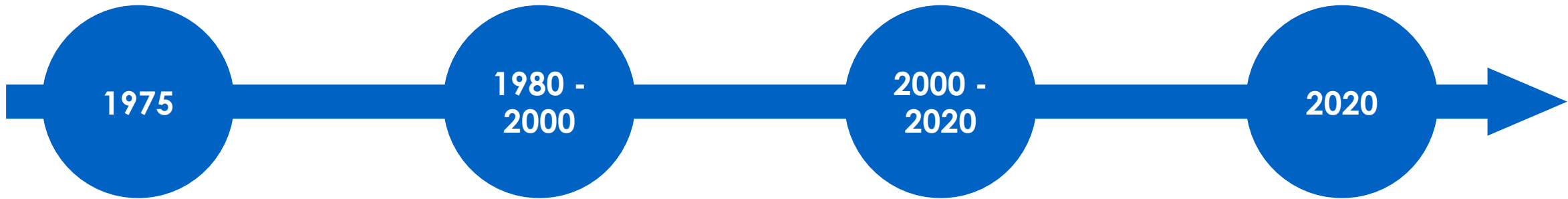
21<sup>st</sup> REGULATORY CONFERENCE

**AMGEN**

August 18, 2023



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1975

First particles AVI machine using **Static Division** technology

1980 - 2000

**Machine Vision** advancements allowed the use of vision inspection for multiple purposes

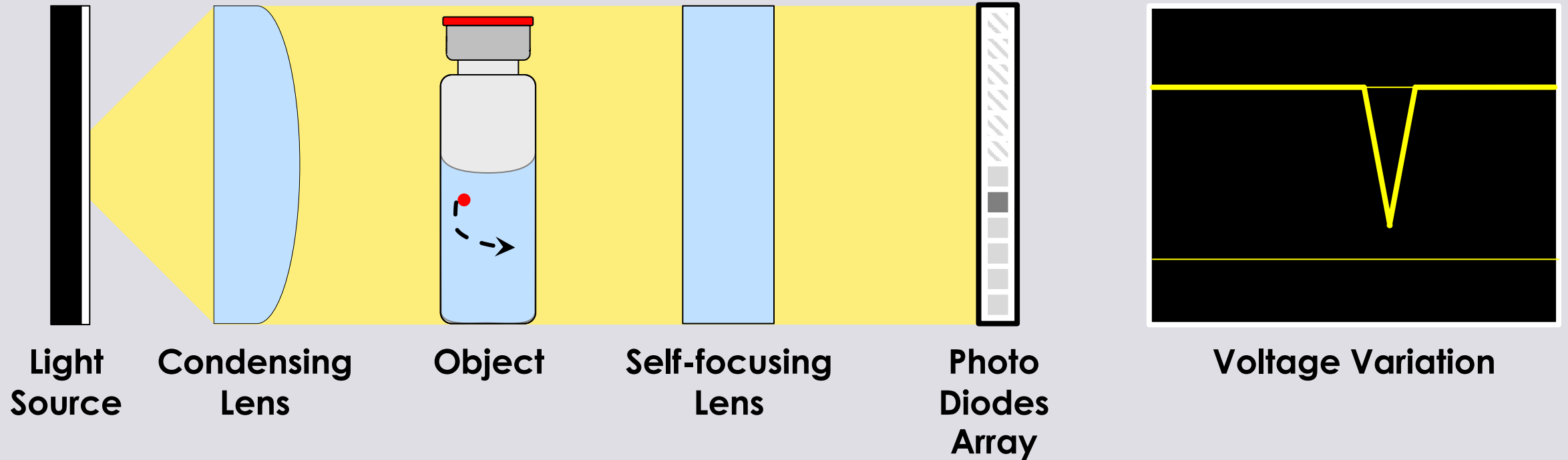
2000 - 2020

Improvements in **Processing Power, Bandwidth** and **Data Storage** allowed the use of Deep Learning in commercial applications

2020

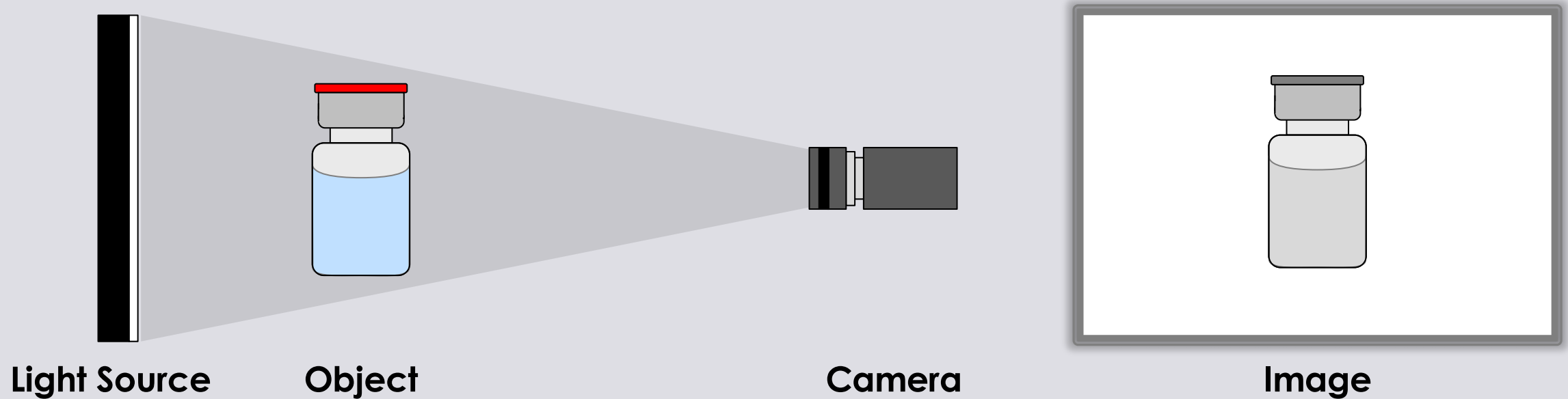
First application of **Deep Learning** for the Automatic Visual Inspection of Parenteral Drug Products

# Static Division Technology



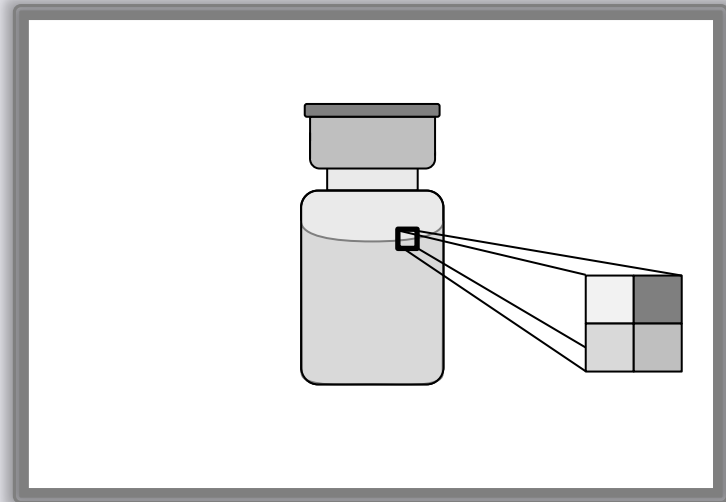
Particles are detected by the variation in voltage created by their shadows against a sensor

# Machine Vision

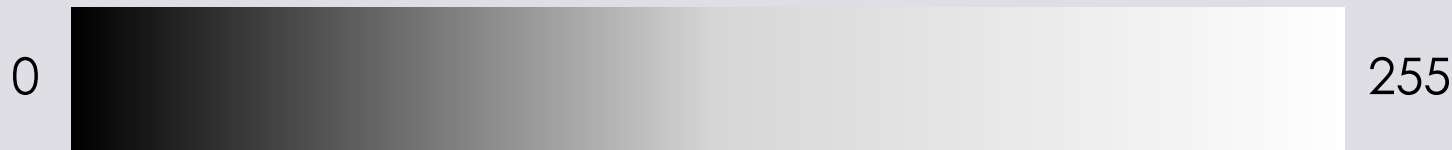


Machine vision is the integration of hardware and software technologies used to **capture**, **process** and **analyze** images

# Basic Image Analysis Using Pixels



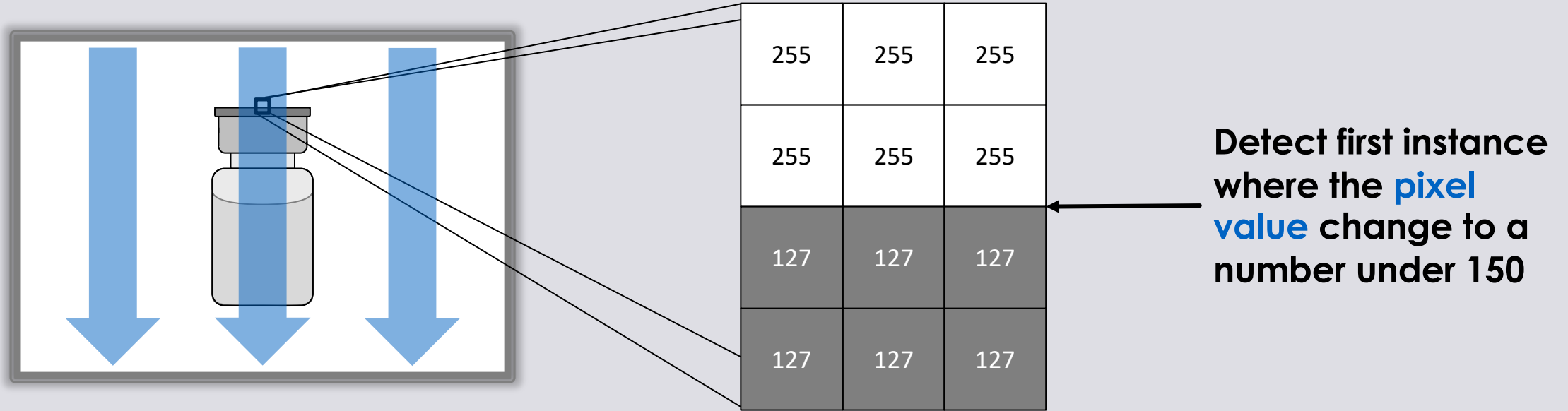
242	127
217	191



Pixel values in a monochromatic image are from 0 to 255

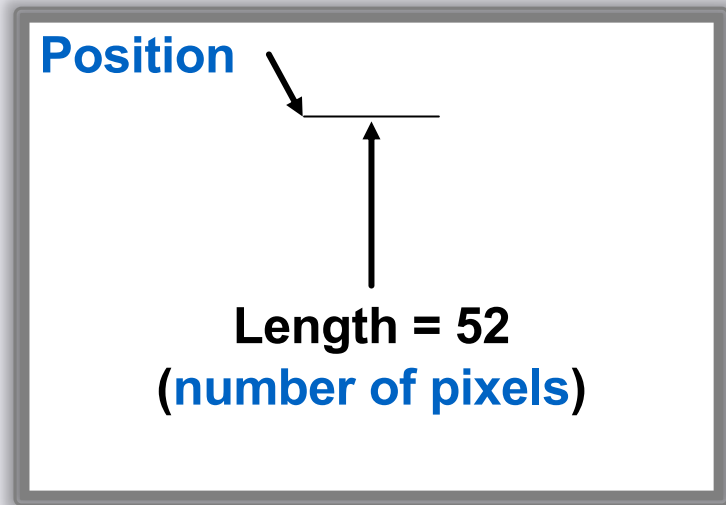
Image analysis is performed using **mathematics** based on pixels  
The **pixel** or picture element is the basic unit of programmable color in a computer image

# Edges are useful tools to detect objects



An Edge in an image is a sharp variation of the intensity or brightness of pixels

# Edges are useful tools to detect objects



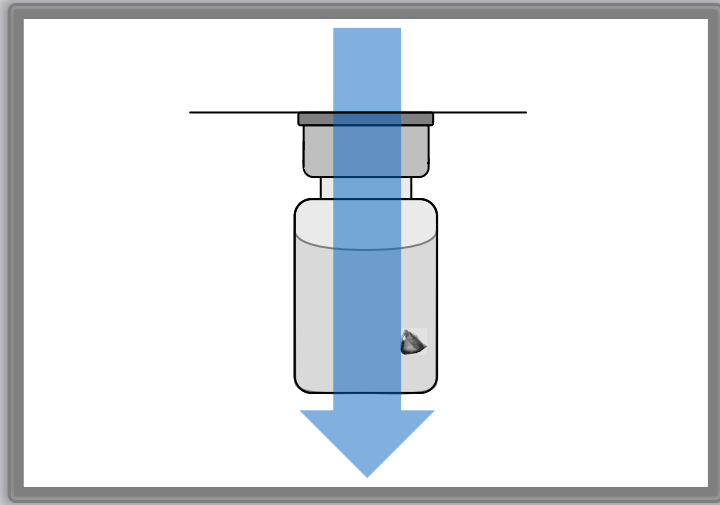
255	255	255
255	255	255
127	127	127
127	127	127

Detect first instance where the **pixel value** change to a number under 150

Multiple mathematical computations can be made using the **pixel values**, **number of pixels** and **position** of pixels

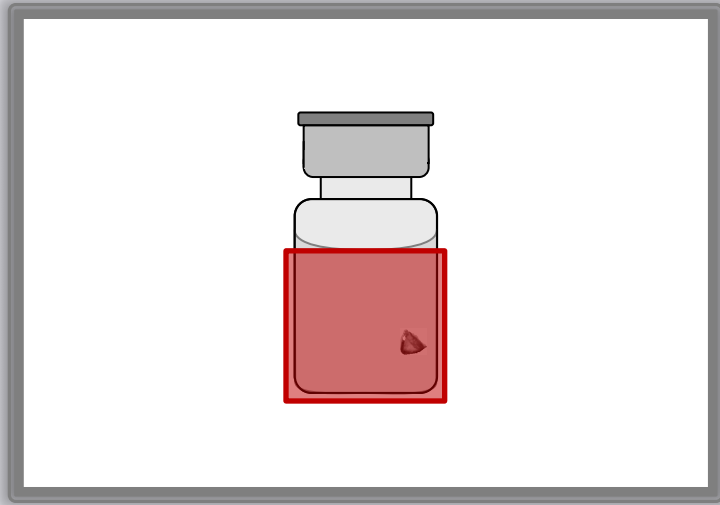


# Particle Detection



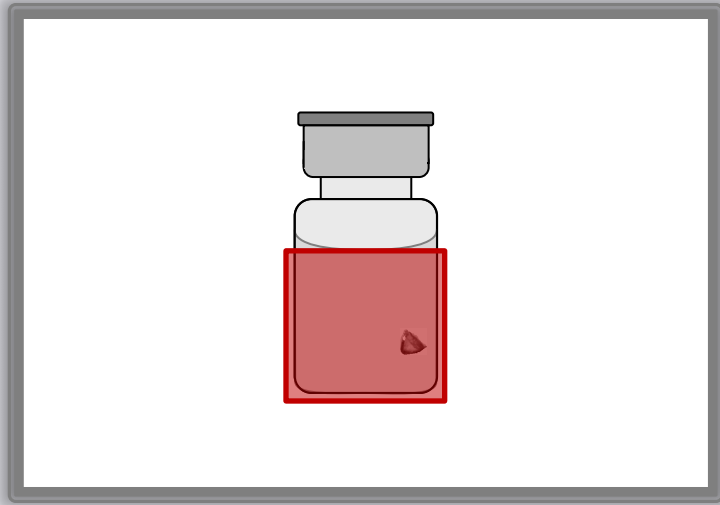
- Identify Object – Edge Tool

# Particle Detection



- Identify Object – Edge Tool
- Select Region of Interest (ROI)

# Particle Detection



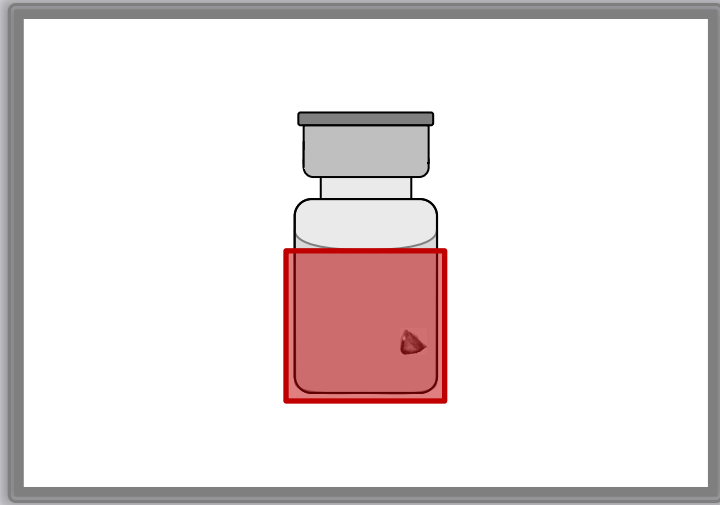
**Glass  
Particle**



A blob is a region of an image in which some pixel properties like intensity or color are approximately constant

- Identify Object – Edge Tool
- Select Region of Interest (ROI)
- Process image - Find Blobs

# Particle Detection



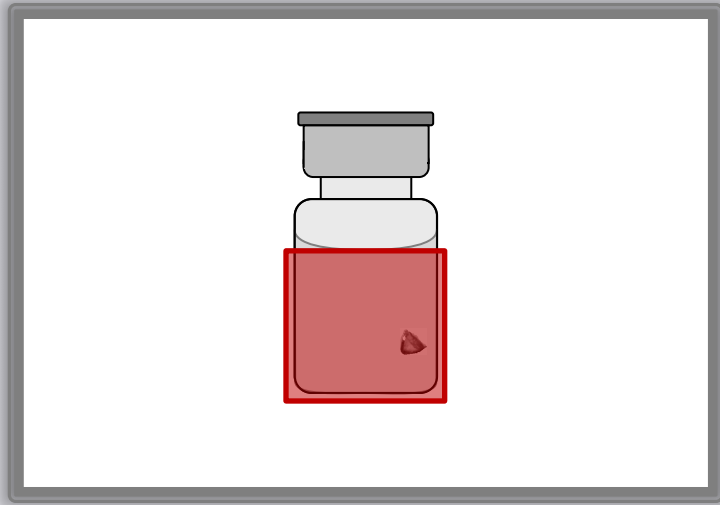
**Glass  
Particle**



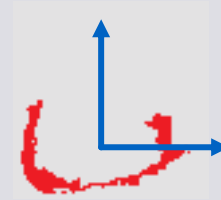
A blob is a region of an image in which some pixel properties like intensity or color are approximately constant

- Identify Object – Edge Tool
- Select Region of Interest (ROI)
- Process image - Find Blobs
- Filter image

# Particle Detection

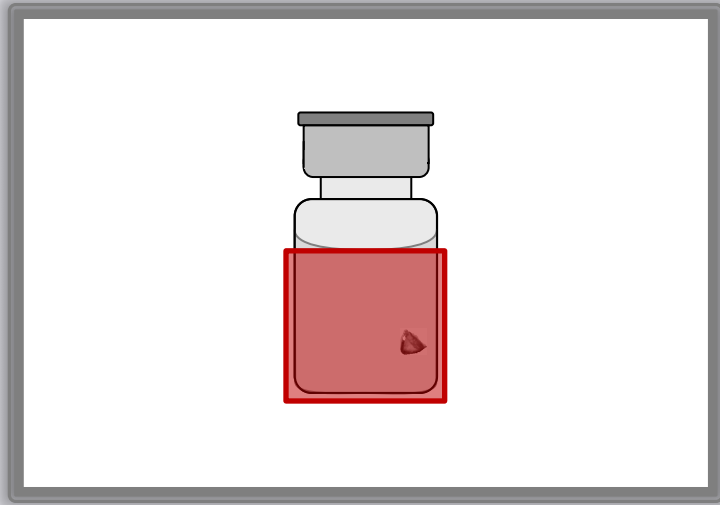


## Glass Particle

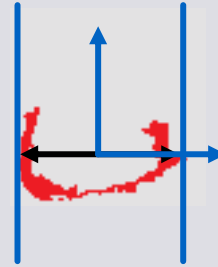


- Identify Object – Edge Tool
- Select Region of Interest (ROI)
- Process image - Find Blobs
- Filter image
- Get Center of Mass

# Particle Detection



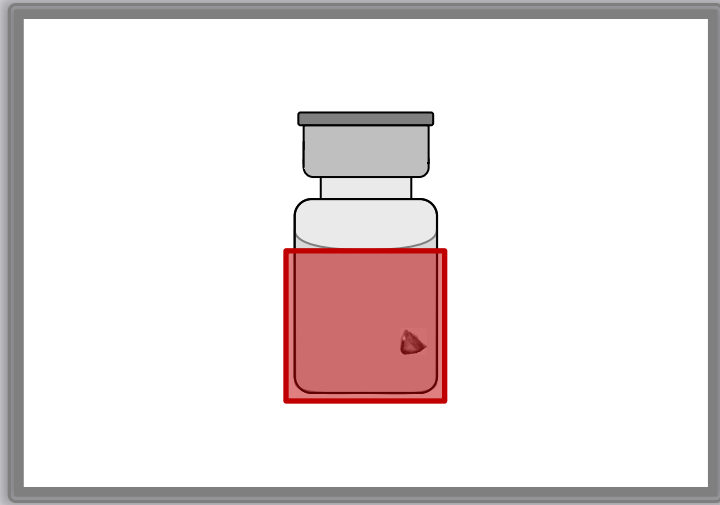
## Glass Particle



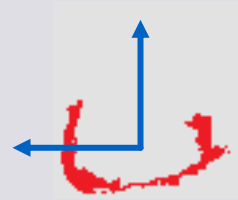
Width = 61

- Identify Object – Edge Tool
- Select Region of Interest (ROI)
- Process image - Find Blobs
- Filter image
- Get Center of Mass
- Calculate Width (number of pixels)

# Particle Detection



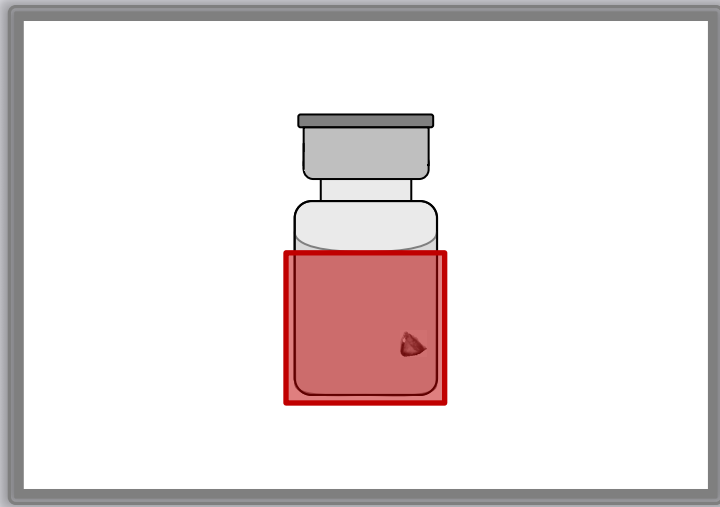
## Glass Particle



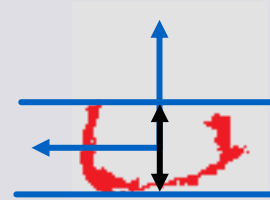
Width = 61

- Identify Object – Edge Tool
- Select Region of Interest (ROI)
- Process image - Find Blobs
- Filter image
- Get Center of Mass
- Calculate Width (number of pixels)
- Rotate Center of Mass Angle

# Particle Detection



## Glass Particle



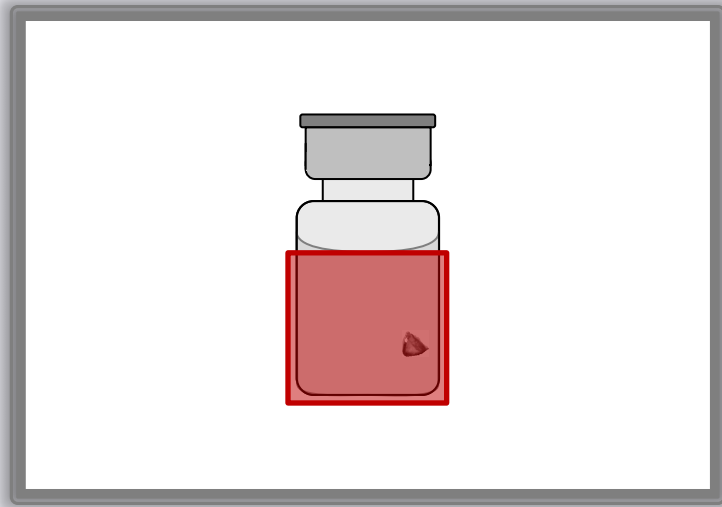
Width = 61

Height = 35

- Identify Object – Edge Tool
- Select Region of Interest (ROI)
- Process image - Find Blobs
- Filter image
- Get Center of Mass
- Calculate Width (number of pixels)
- Rotate Center of Mass Angle
- Calculate High (number of pixels)



# Particle Detection



## Glass Particle



Width = 61

Height = 35

Ratio = Width ÷ Height

Ratio = 1.74

Ratio ≠ 1.00

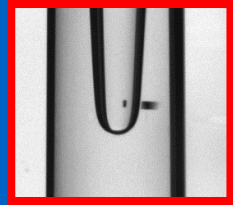
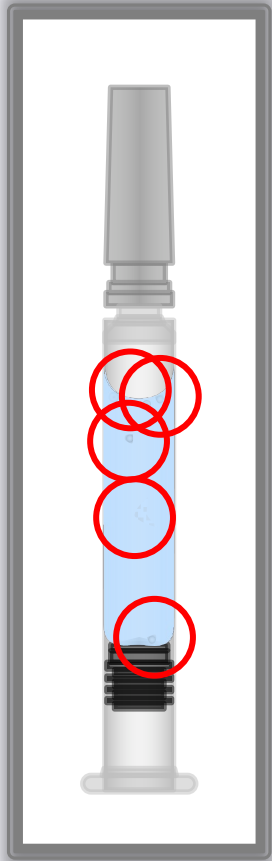
- Identify Object – Edge Tool
- Select Region of Interest (ROI)
- Process image - Find Blobs
- Filter image
- Get Center of Mass
- Calculate Width (number of pixels)
- Rotate Center of Mass Angle
- Calculate High (number of pixels)
- If ratio ≠ 1, consider it as a particle

# Current Automatic Visual Inspection (AVI) Technologies for Particle Detection

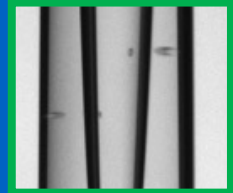
- **Spin** units to promote particles movement and dissipate bubbles
- **Tracking system** to detect movement, identify particles and differentiate them from bubbles
- **Static Division** use light transmission method and a diode array to detect light reduction (shadows) caused by moving particles in solution blocking light
- **Image subtraction** that compares different images of the same static unit and identify changes in the images
- **Polarized filters** to highlight particle features

Even with multiple techniques, it is still a challenge to effectively detect particles

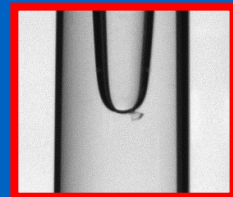
# Deep Learning can help overcome some challenges in Automatic Visual Inspection



**Plastic**



**Bubbles**



**Glass**

- Small containers do not promote the movement of particles, i.e., small syringes
- Particles that adhere to the meniscus are difficult to detect
- Bubbles inherent to the formulation of the product that are similar to particles
- Glass particles in solution similar to bubbles
- Heavy particles allocated on top of the plunger
- Protein aggregates inherent to the product are not defects, but are similar to particles
- Viscous or non-clear products

# Deep Learning is a Subset of Machine Learning, Which is a Subset of Artificial Intelligence

## Artificial Intelligence (AI)

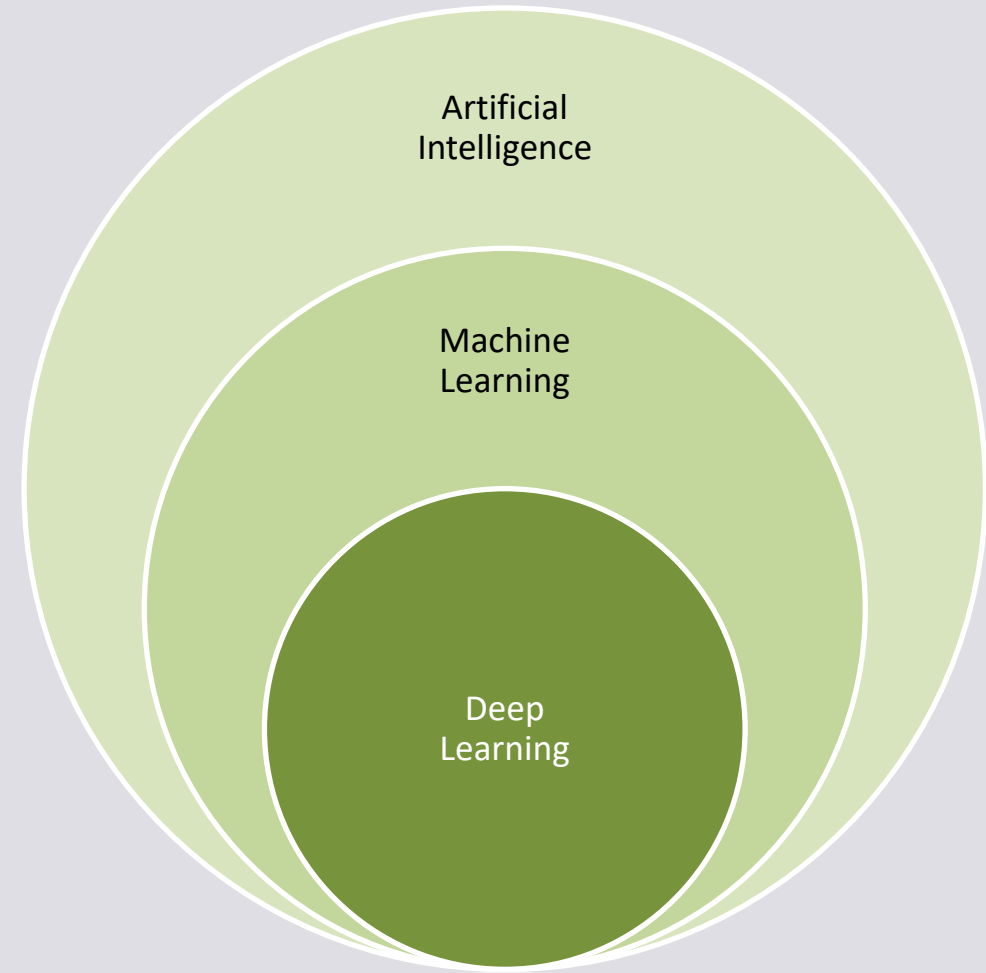
AI involves machines that can perform tasks that are **characteristic of human intelligence**

## Machine Learning (ML)

A subset of AI which gives the computer the ability to **learn without being explicitly programmed**

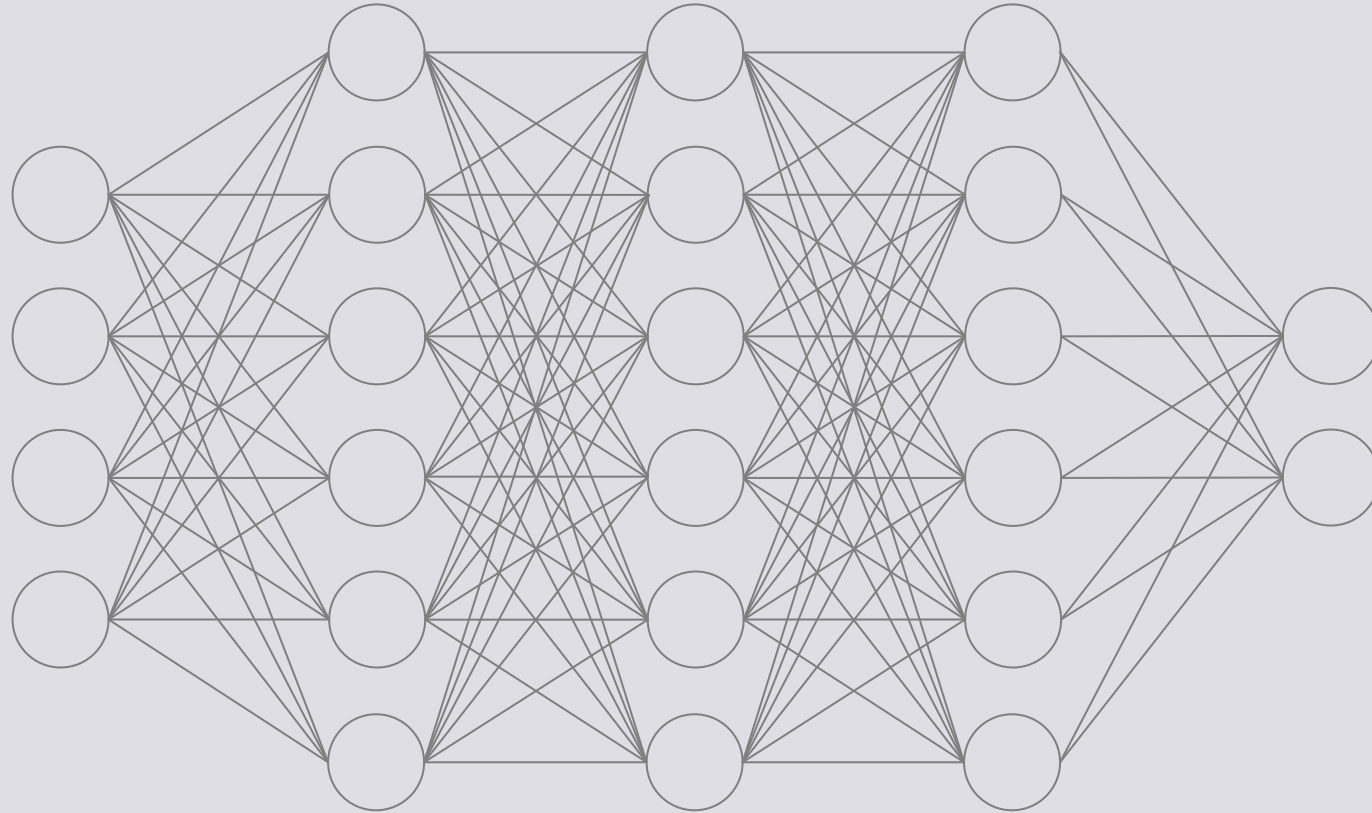
## Deep Learning (DL)

A subset of ML inspired by the structure and the function of the brain which use **Artificial Neural Networks** to find patterns in data and to infer the outcome of something it has never been exposed before



# Artificial Neural Network

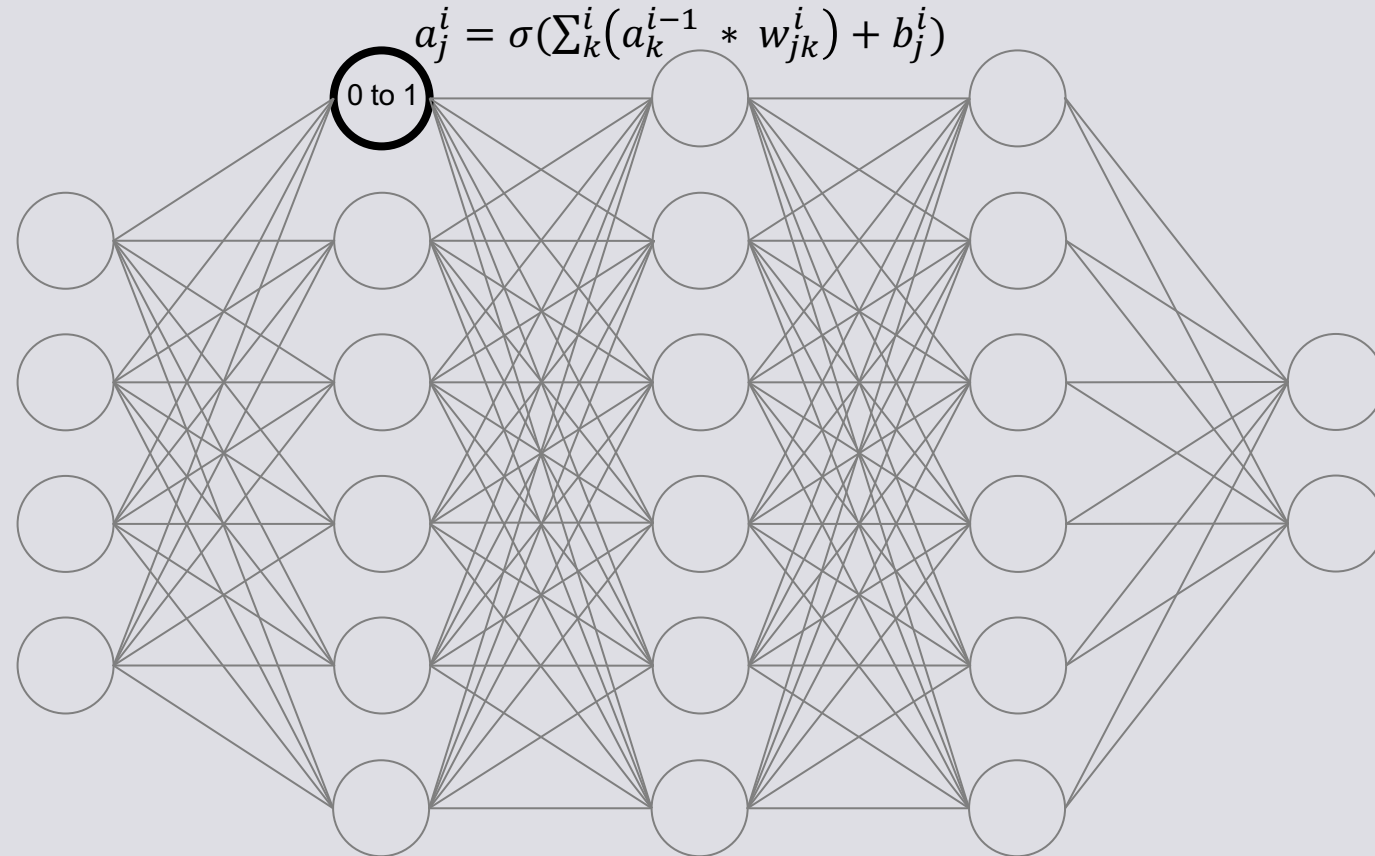
Computing system based on interconnected neurons or nodes (circles) arranged in layers



# Artificial Neural Network

Computing system based on interconnected neurons or nodes (circles) arranged in layers

Each neuron is a mathematical equation that use previous neurons and weight values as input

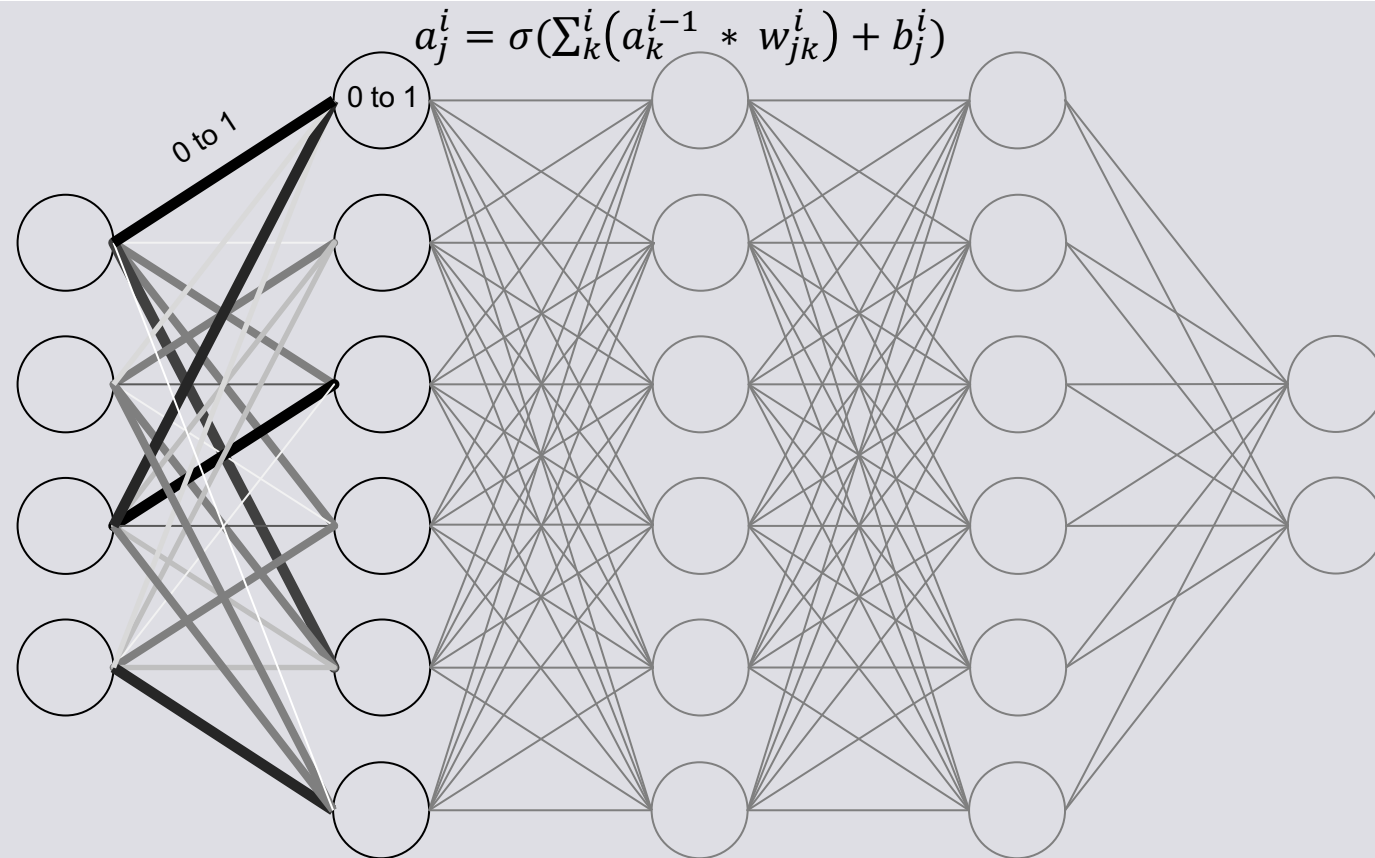


# Artificial Neural Network

Computing system based on interconnected neurons or nodes (circles) arranged in layers

Each neuron is a mathematical equation that use previous neurons and weight values as input

Each connection or line is a weight value indicating how important is the relation between neurons

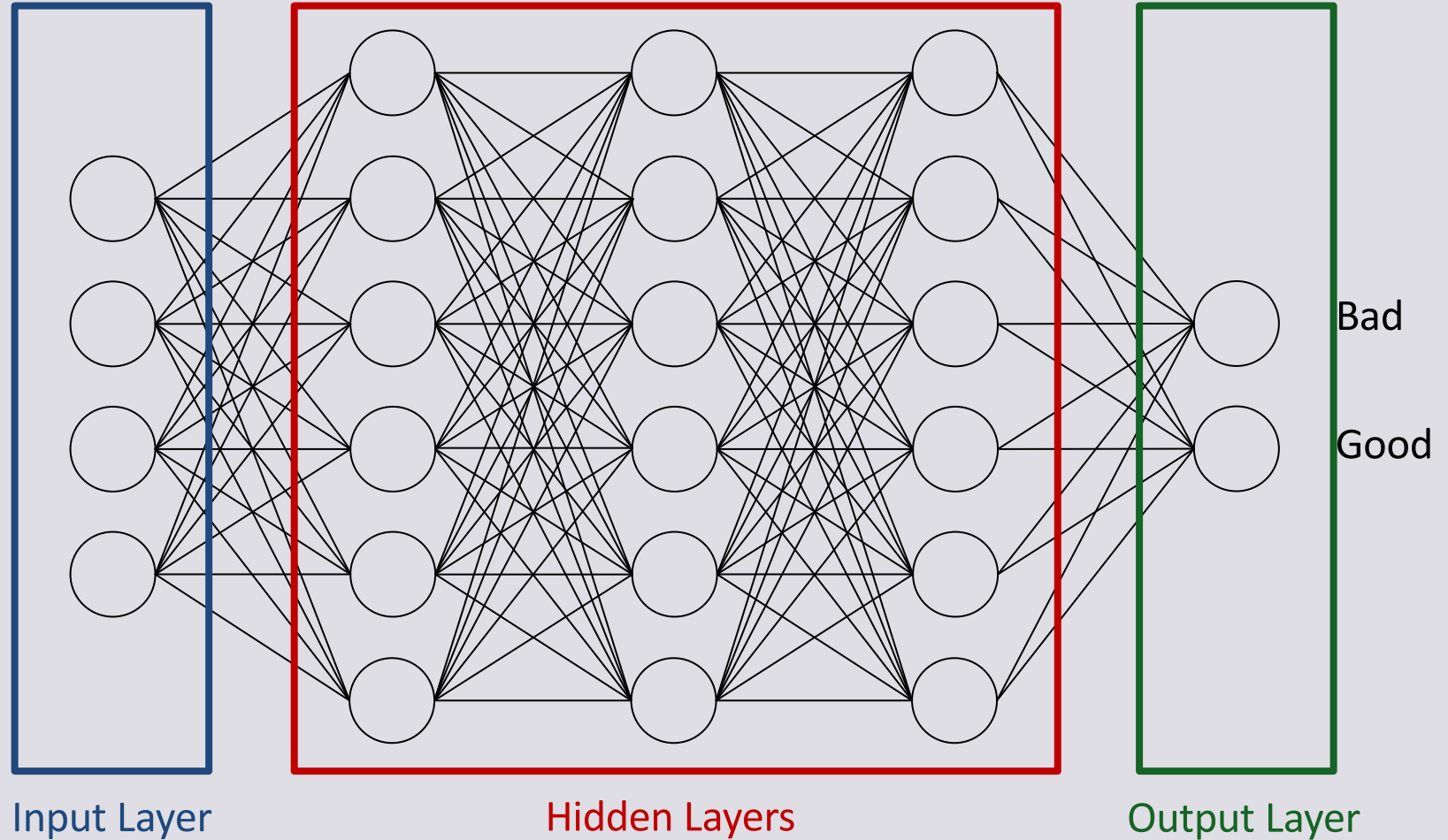


# Artificial Neural Network Structure

Input Layer – Input from the physical world

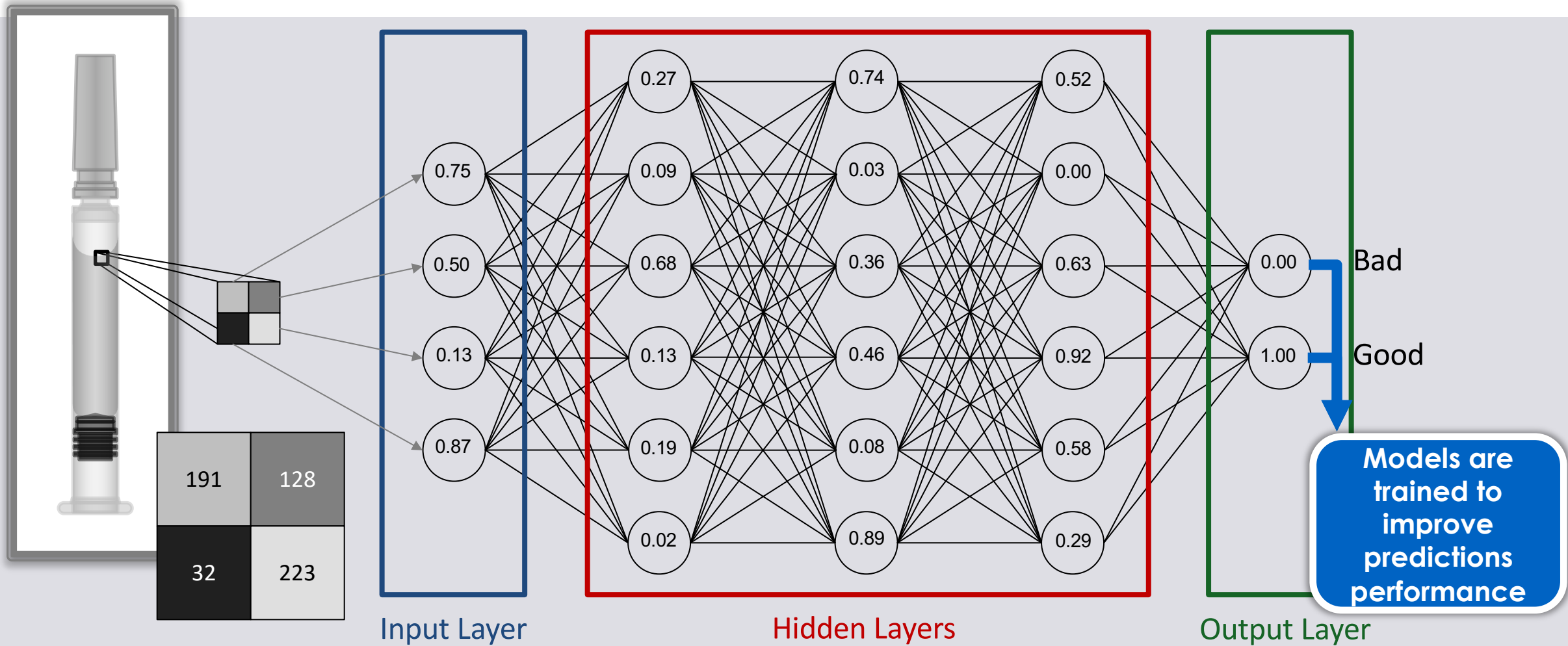
Hidden Layers – used for processing the inputs through activation functions

Output Layer – Contains the result of the inference made by the model

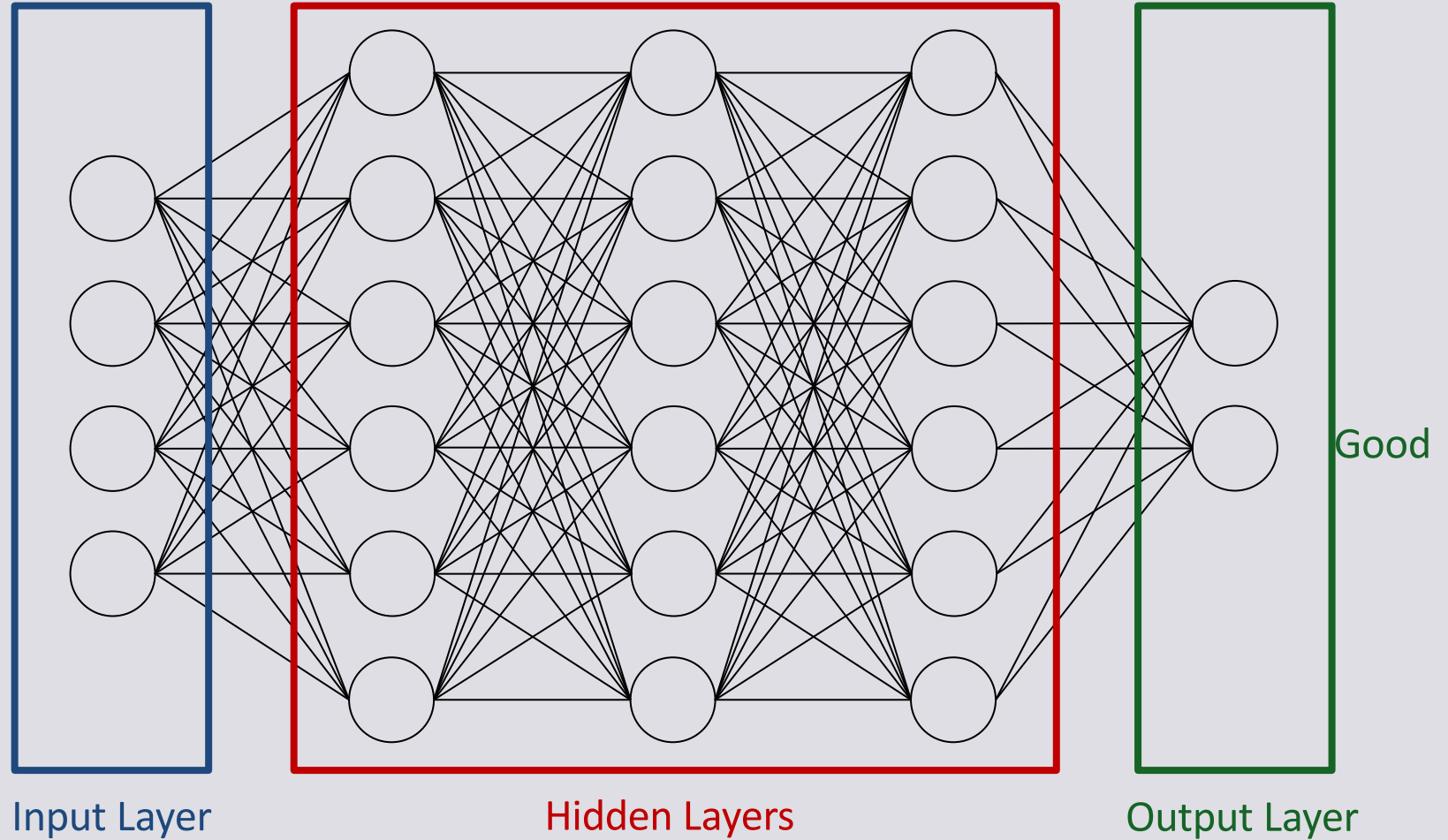
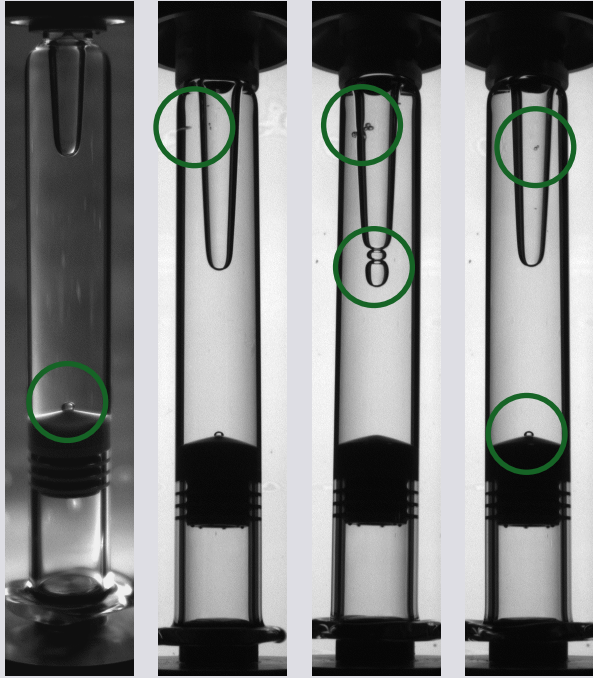




# Artificial Neural Network Prediction

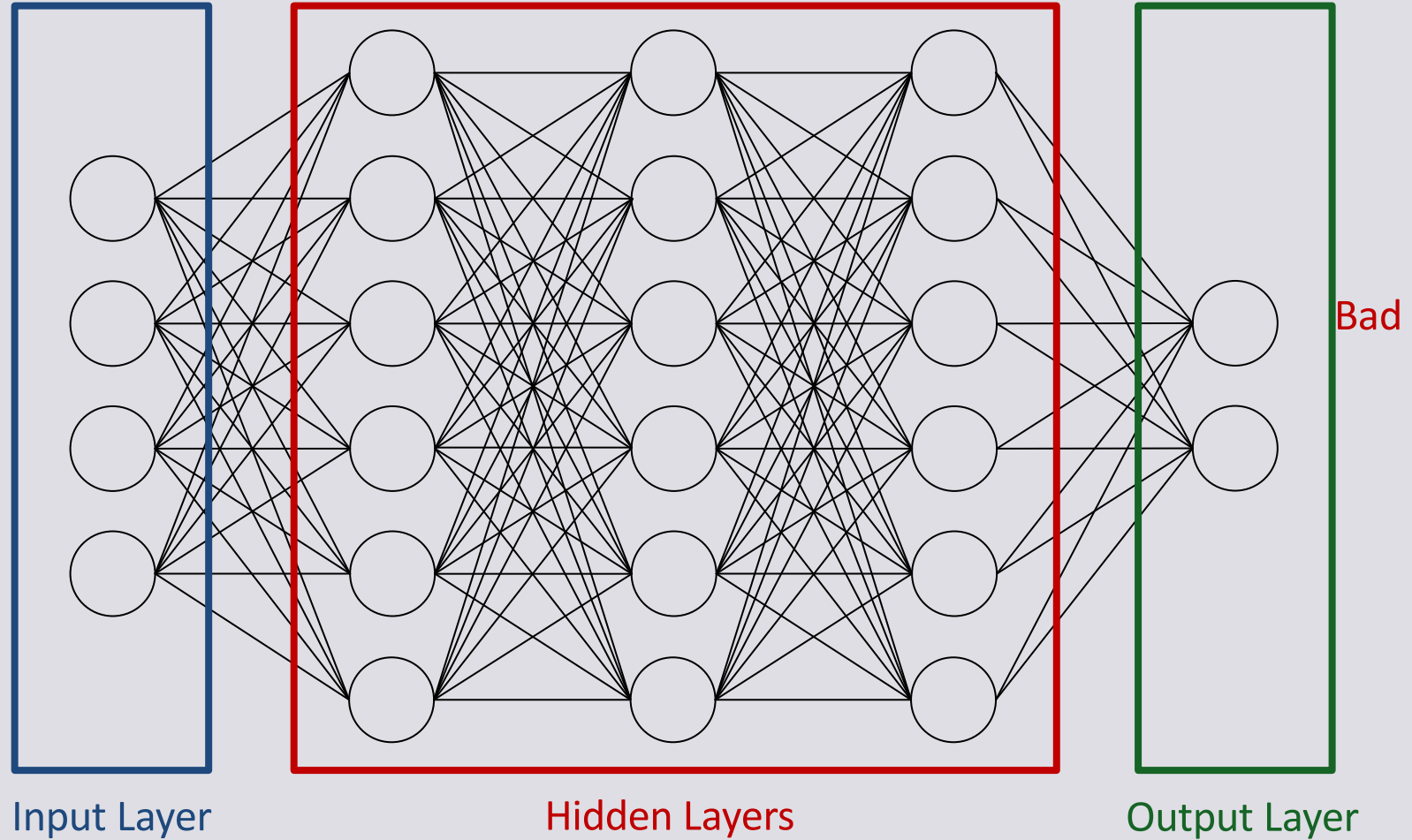
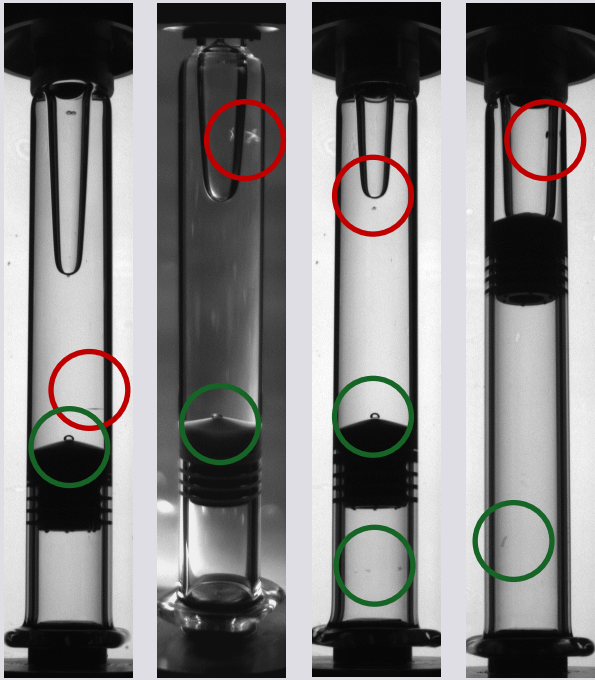


# Artificial Neural Network - Training - Good



\*Test units made in the laboratory to simulate defects

# Artificial Neural Network - Training - **Bad**



\*Test units made in the laboratory to simulate defects

# Model Testing – Confusion Matrix

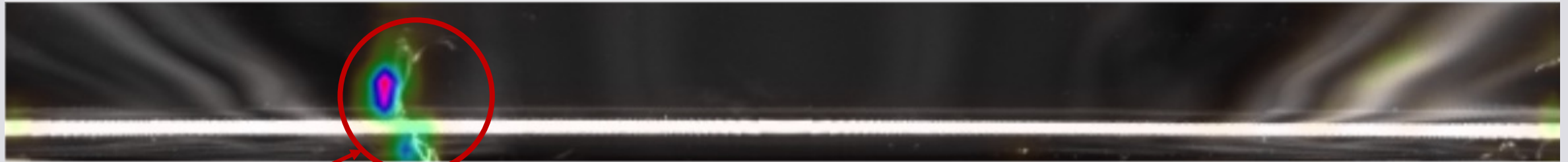
A confusion matrix is a tool used to measure the performance of a Deep Learning model by summarizing the number of correct and incorrect predictions made

		Predicted (as inferred by the model)	
		Bad	Good
Actual (as labeled)	Bad	True Bad	False Good (Type I Error)
	Good	False Bad (Type II Error)	True Good

Detection Rate = Predicted Bad ÷ Actual Bad

False Fails = Predicted Bad ÷ Actual Good

# Deep Learning Heat Map



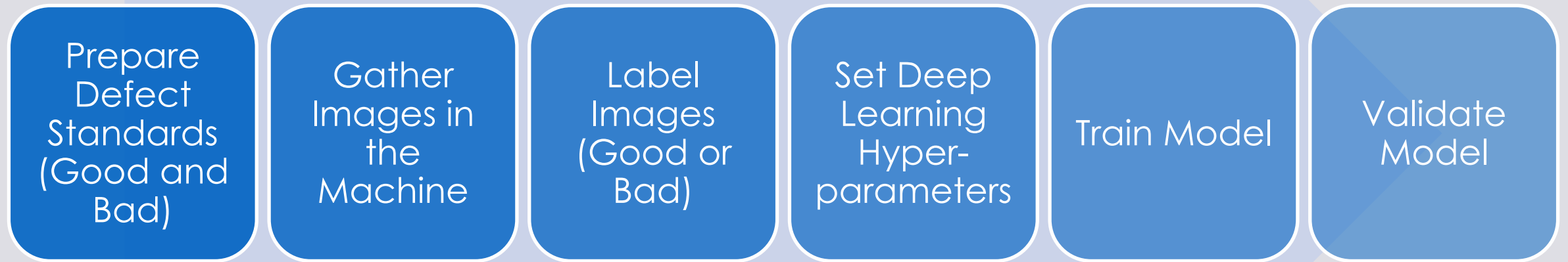
Crack

Original Image

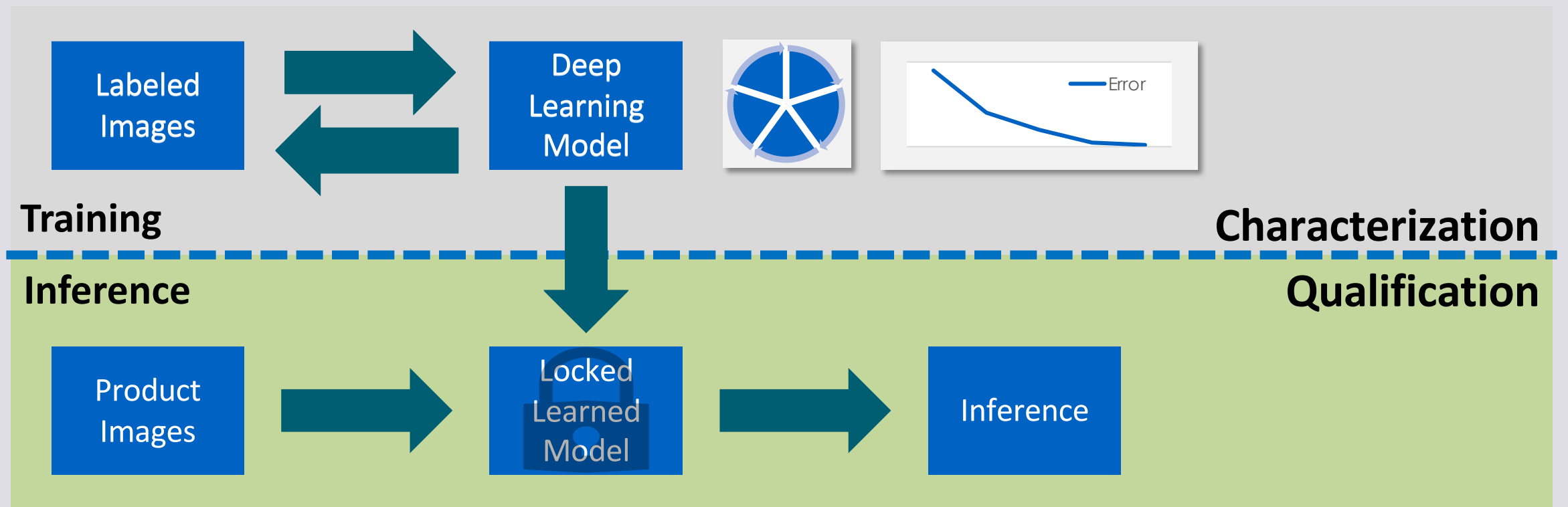
Heatmap

Heat maps are used to highlight the areas in the image which influenced the neural network in the inference made

# Deep Learning Model Generation

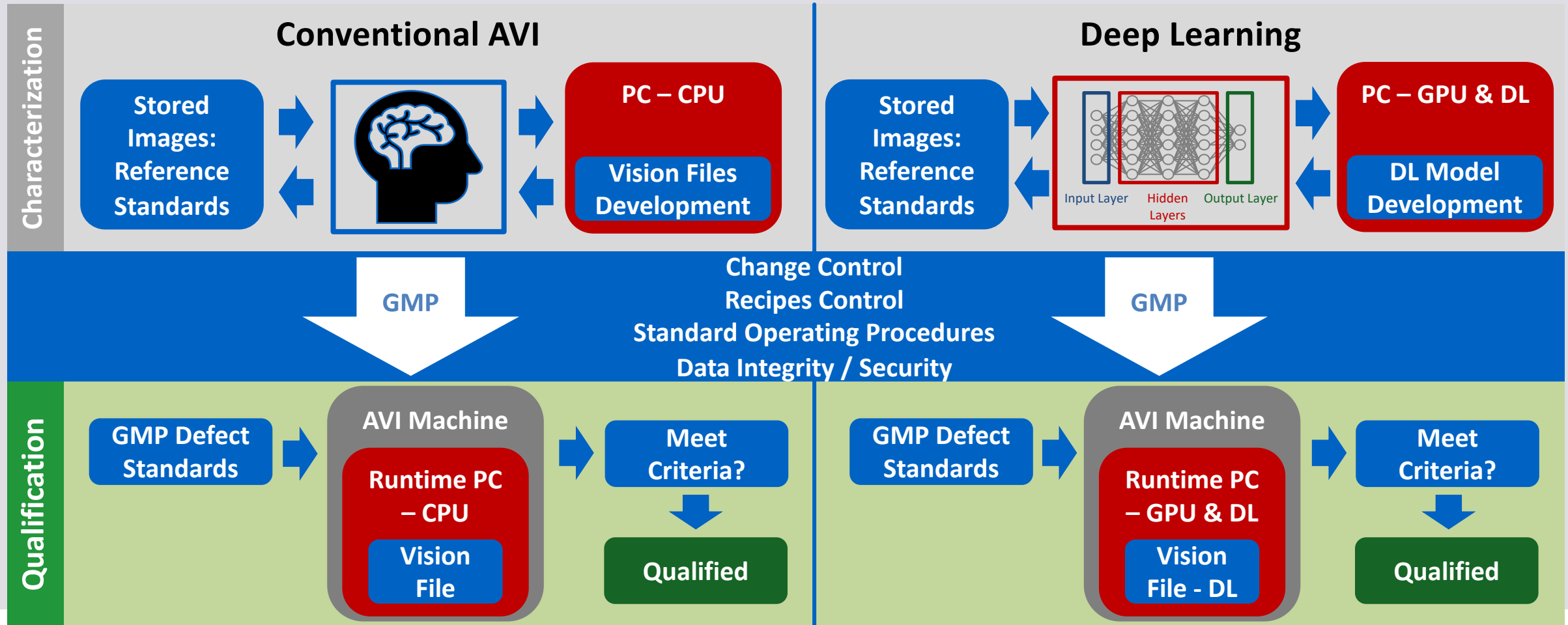


# Deep Learning Characterization and Qualification



Once a model is locked, there are no changes in the Neural Network  
If 1 image is evaluated 1 million times, 1 million times we will obtain the **same result**

# AVI Qualification Approach

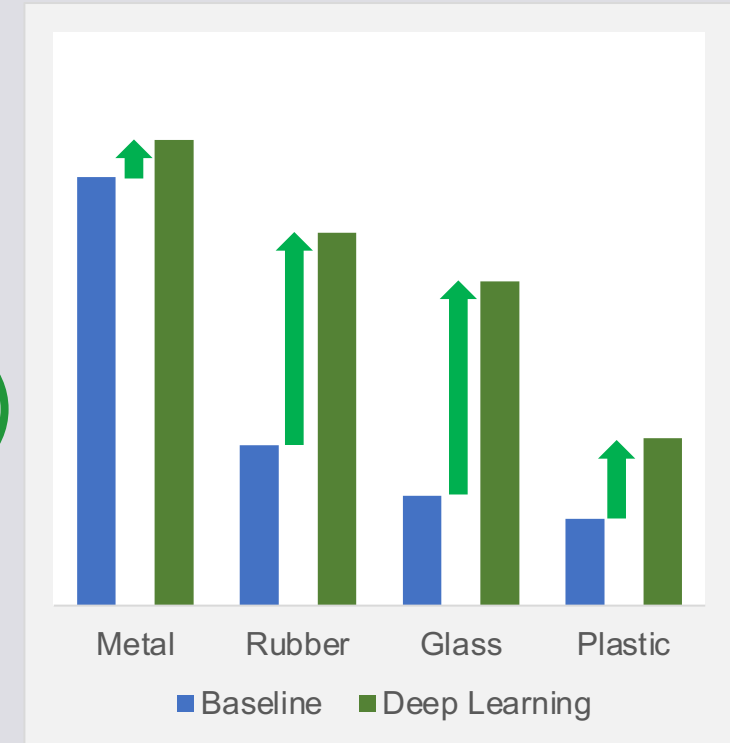
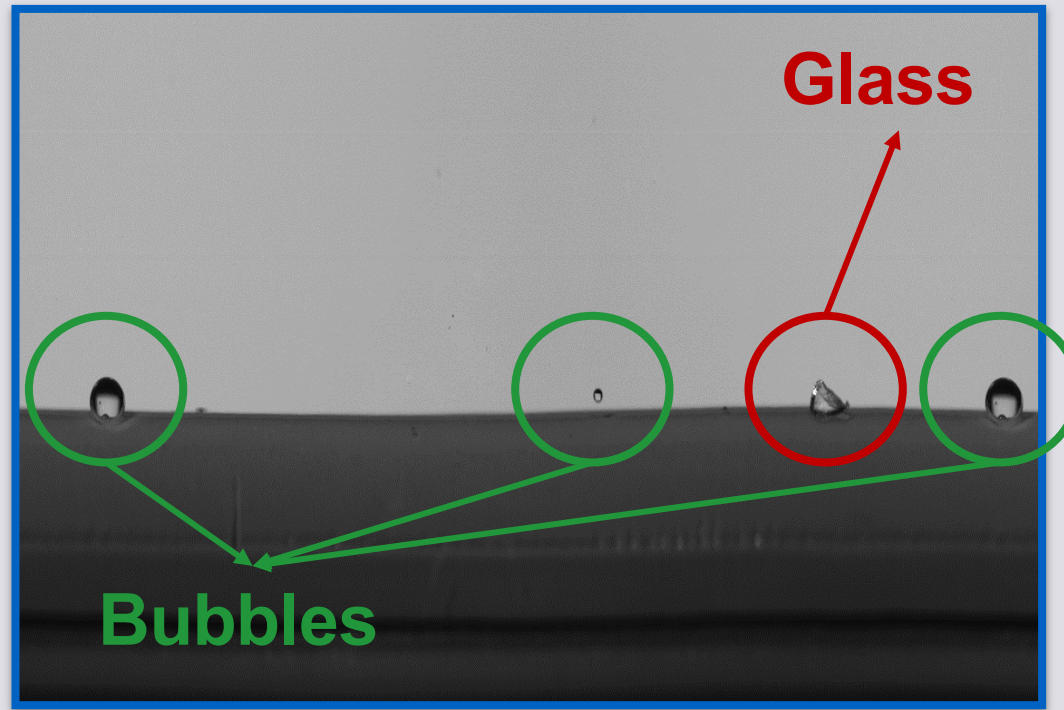
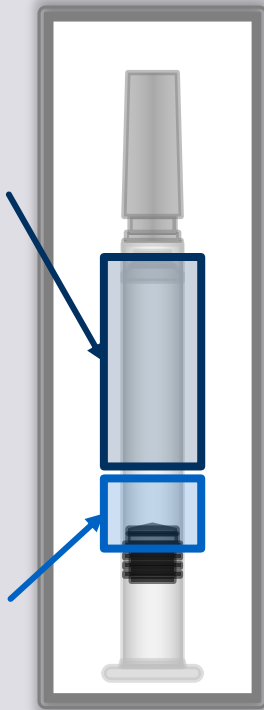




# Studies have shown an increase in detectability of particles in syringes

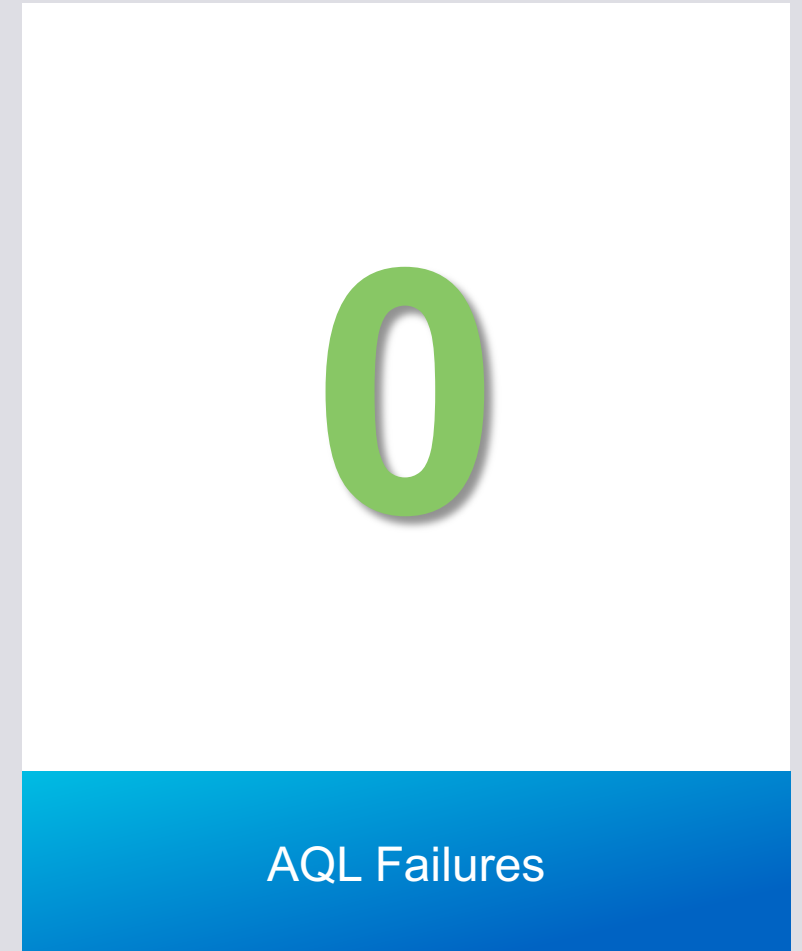
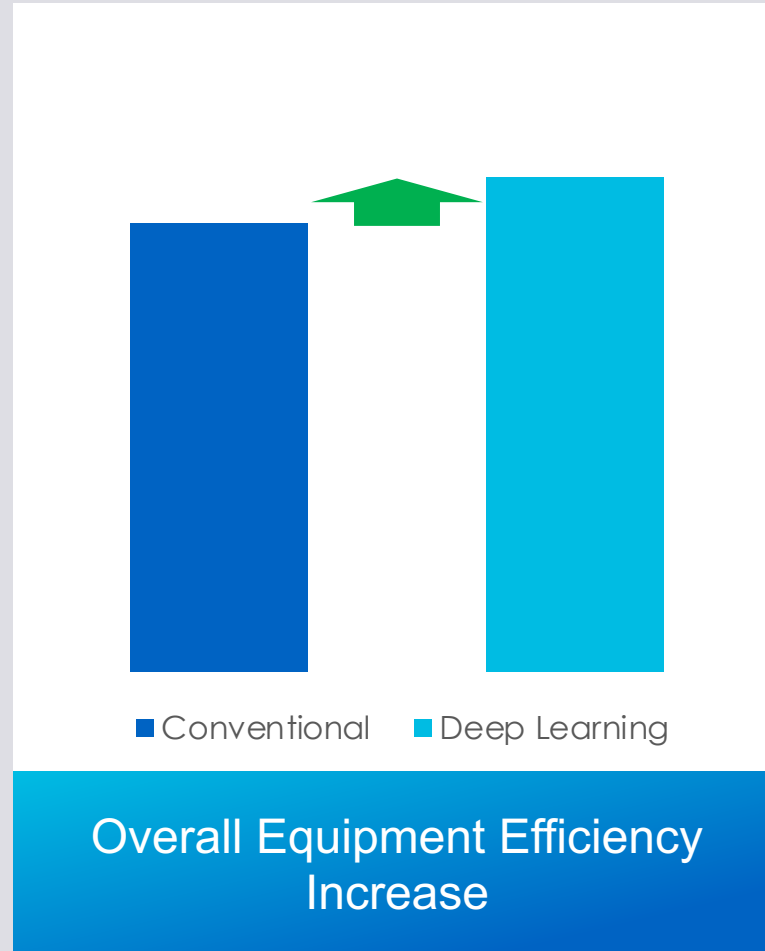
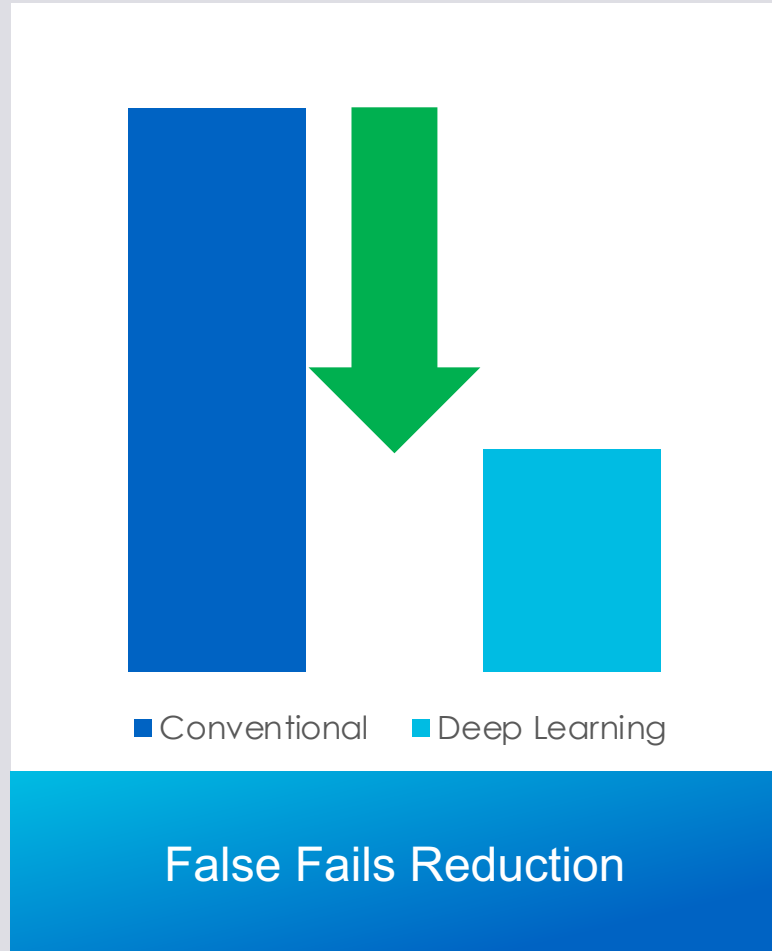
Region Inspected By Other Stations

Inspection Region



Machine vision is the integration of hardware and software technologies used to capture, process and analyze images

# Results Obtained After Millions of Units Processed with Deep Learning for the Inspection of Pre-Filled Syringes



# Deep Learning models require a significant number of images which brings several challenges

- Line time required for image gathering
- Cost of test material and components
- Availability of test material and components
- Resources time for image labeling

Many of these challenges can be overcome with technology

# Work have been done to augment images using Generative Adversarial Networks

## Deep Learning Image Augmentation using Inpainting with Partial Convolution and GANs

by Aik Jun Tan

Submitted to the Sloan School of Management and Department of Electrical Engineering and Computer Science in partial fulfilment of the requirements for the degree of

**Master of Business Administration and Master of Science in Electrical Engineering and Computer Science** in conjunction with the Leaders in Global Operations (LGO) program at the

MASSACHUSETTS INSTITUTE OF TECHNOLOGY

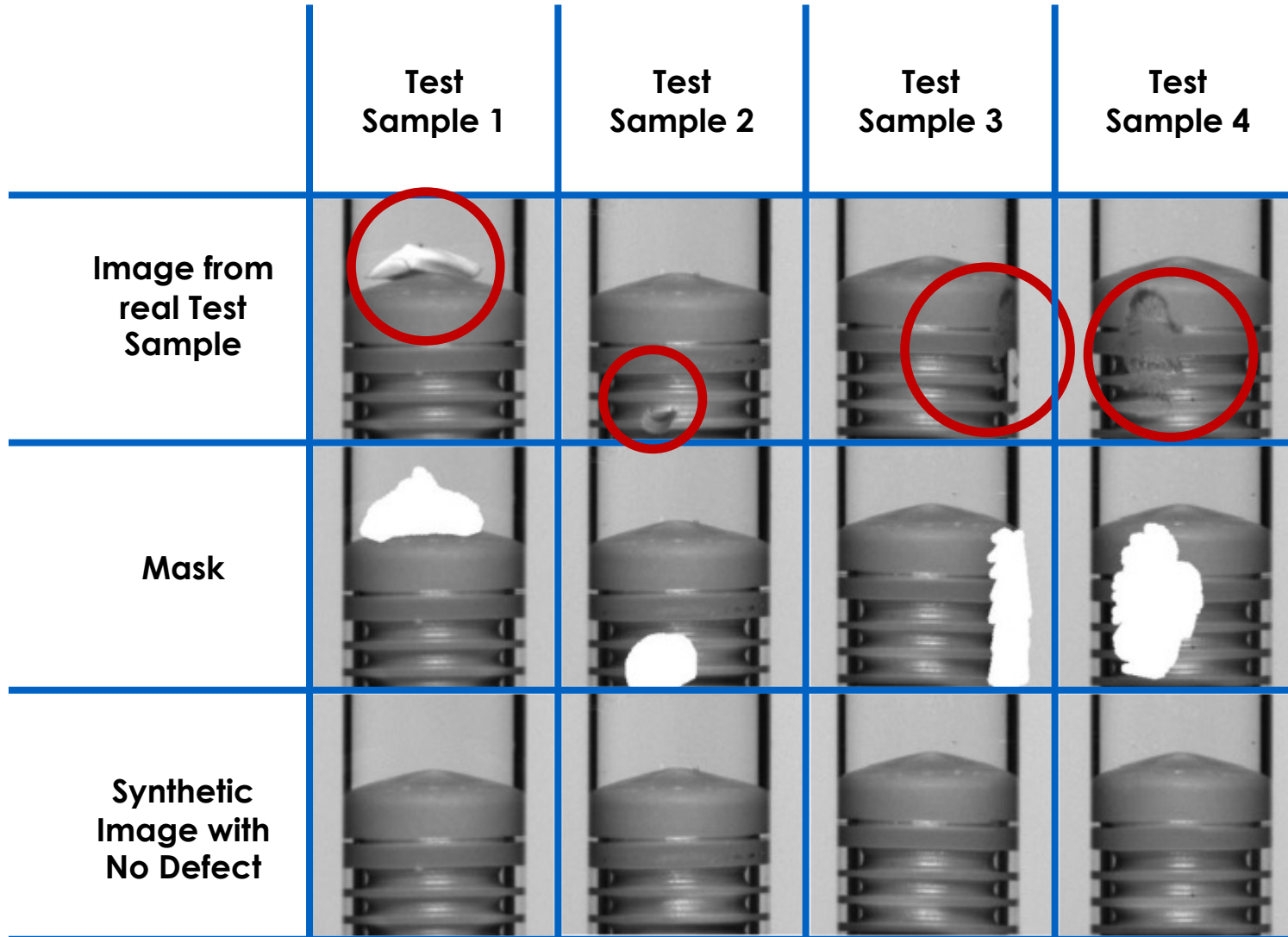
June 2021

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This tool can be used to provide a baseline model that can be further fine tuned

# Stopper Defects Removed



# Syringe Defects Removed

Test Sample 1

Test Sample 2

Image from  
real Test  
Sample



Synthetic  
Image with  
No Defect

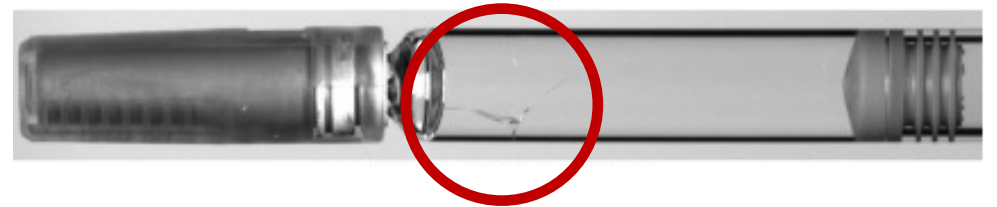


# Syringe Defects Removed

Test Sample 3

Test Sample 4


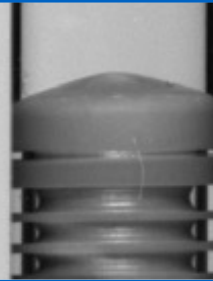





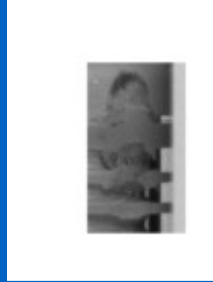

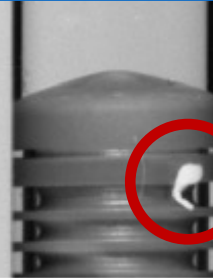


Image from  
real Test  
Sample



Synthetic  
Image with  
No Defect



# Stopper Defects Added

	Test Sample 1	Test Sample 2	Test Sample 3	Test Sample 4
Image from real Test Sample				
Mask				
Synthetic Image with No Defect				



# Syringe Defects Added

Test Sample 1

Test Sample 2

Test Sample 3

Image from  
real Test  
Sample



Synthetic  
Image with  
No Defect



# Possibilities in the Future of AVI

- Increased use of Deep Learning for the inspection of parenteral drug product
- Image augmentation will reduce development time and implementation cost
- Image gathering will allow for continuous improvements with periodic fine tuning of deep learning models
- Vendors will provide machines with little to no need for development
- AVI will be feasible for High Mix Low Volume products
- Technology will allow more inspections at earliest stages of the process, detecting weak signals while processing a batch
- AVI systems will have the capability to classify particle types, facilitating the identification of potential root causes and their resolution

# THANK YOU

