



# ■ Deep Learning for Computer Vision

Program Magister Terapan PENS



# Applications of Deep Learning for Computer Vision

# 1. Image Classification

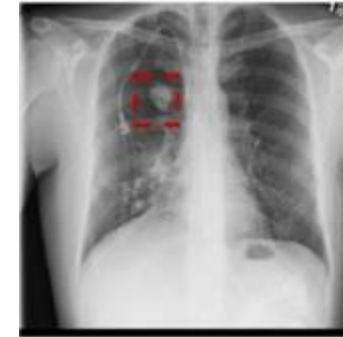
- Assigning a label to an entire image or photograph.

- binary classification:**

- Labeling an x-ray as cancer or not.
- Labelling a photograph as cat or not

- multiclass classification:**

- Classifying a handwritten digit.
- Assigning a name to a photograph of a face.



Cancer or not ?



5 ?

## Classification

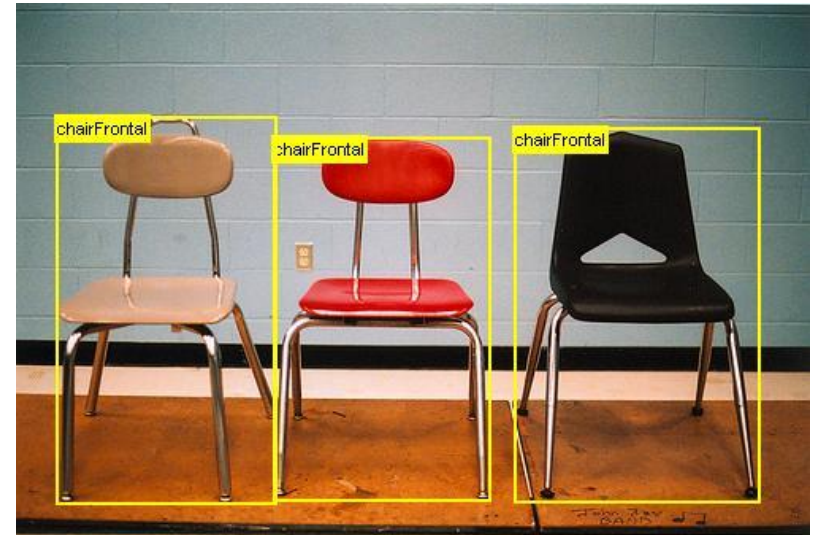


CAT

Example of Handwritten Digits From the MNIST Dataset

## 2. Image Classification with Localization

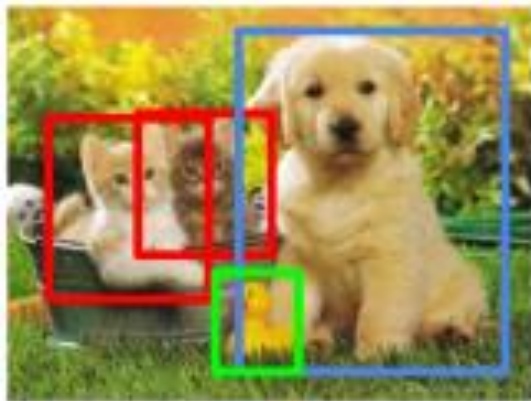
- Assigning a class label to an image and showing the location of the object in the image by a bounding box (drawing a box around the object).
  - Classifying photographs of animals and drawing a box around the animal in each scene.
  - Labeling an x-ray as cancer or not and drawing a box around the cancerous region.
- Datasets : PASCAL VOC (VOC 2012) and ILSVRC2016 (comprised of 150,000 photographs with 1,000 categories of objects).



Example of Image Classification With Localization PASCAL VOC 2012

# 3. Object Detection

- Label an image that may contain multiple objects that require localization and classification.
- It is more complex than simple image classification or image classification with localization, as often there are **multiple objects** in the image of different types.
- Common datasets: PASCAL VOC (VOC 2012) and MS COCO.



CAT, DOG, DUCK



Example of Object Detection with Faster R-CNN on the MS COCO Dataset

## 4. Object Segmentation

- Or **semantic segmentation**, is the task of **object detection** where a line is drawn around each object detected in the image.
- Identifies the **specific pixels** in the image that belong to the object.
- Common datasets: KITTI, PASCAL VOC (VOC 2012) and MS COCO.



CAT, DOG, DUCK



Example of Object Segmentation on the COCO Dataset Taken from “Mask R-CNN”.

# 5. Style Transfer

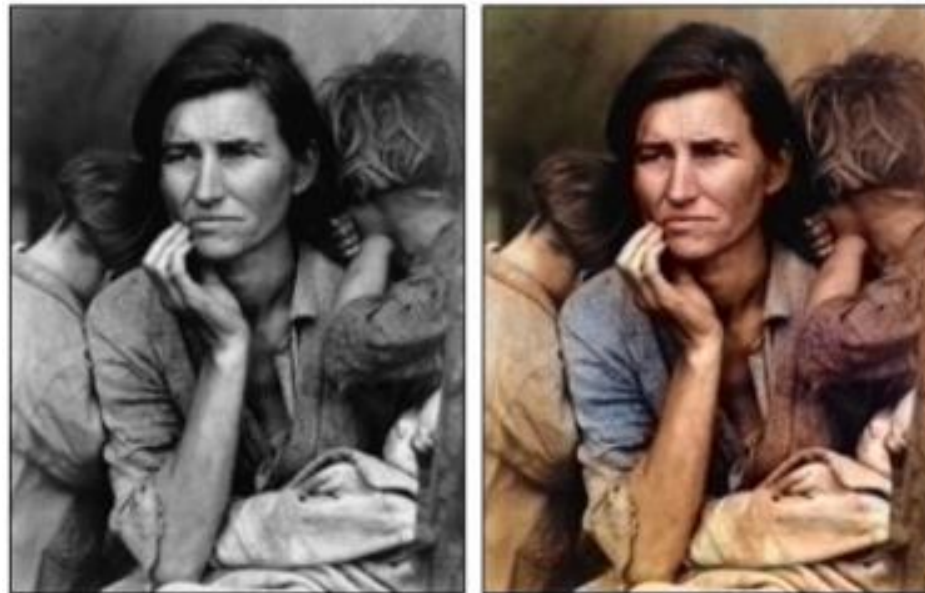
- Style transfer is the task of **learning style from one or more images and applying that style to a new image**.
- Example: applying the style of specific famous artworks (e.g. by Pablo Picasso or Vincent van Gogh) to new photographs.
- Dataset: using **famous artworks** and photographs from standard computer vision datasets.
- May not have an objective evaluation



Example of Neural Style Transfer From Famous Artworks to a Photograph. Taken from “A Neural Algorithm of Artistic Style”

# 6. Image Colorization

- Image colorization involves **converting a grayscale image to a full color image**.
- Examples: colorizing **old black and white photographs and movies**.
- May not have an objective evaluation.
- Datasets: **using existing photo** datasets and **creating grayscale versions** of photos that models must learn to colorize.



Examples of Photo Colorization Taken from “Colorful Image Colorization”



# 7. Image Reconstruction/Restoration

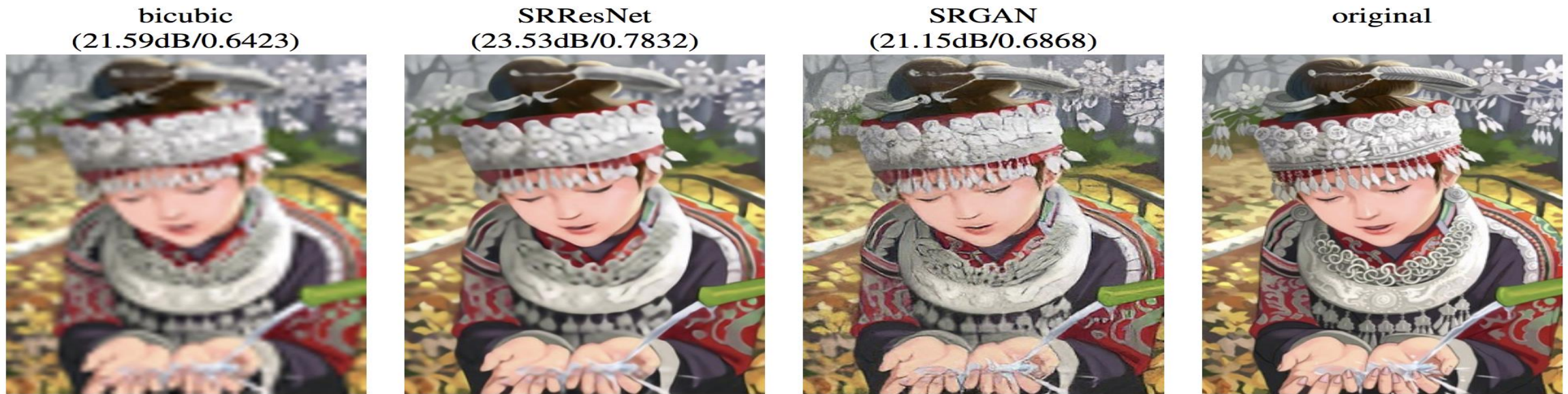
- It is the task of **filling in missing or corrupt parts** of an image.
- Example: reconstructing old, damaged black and white photographs and movies.
- May not have an objective evaluation
- Datasets: **using existing photo** datasets and **creating corrupted versions** of photos that models must learn to repair.



Example of Photo Inpainting. Taken from “Image Inpainting for Irregular Holes Using Partial Convolutions”

# 8. Image Super-Resolution

- It is the task of **generating a new version of an image with a higher resolution and detail** than the original image.
- Datasets: **using existing photo** datasets and **creating down-scaled versions** of photos or which models must learn to create super-resolution versions.

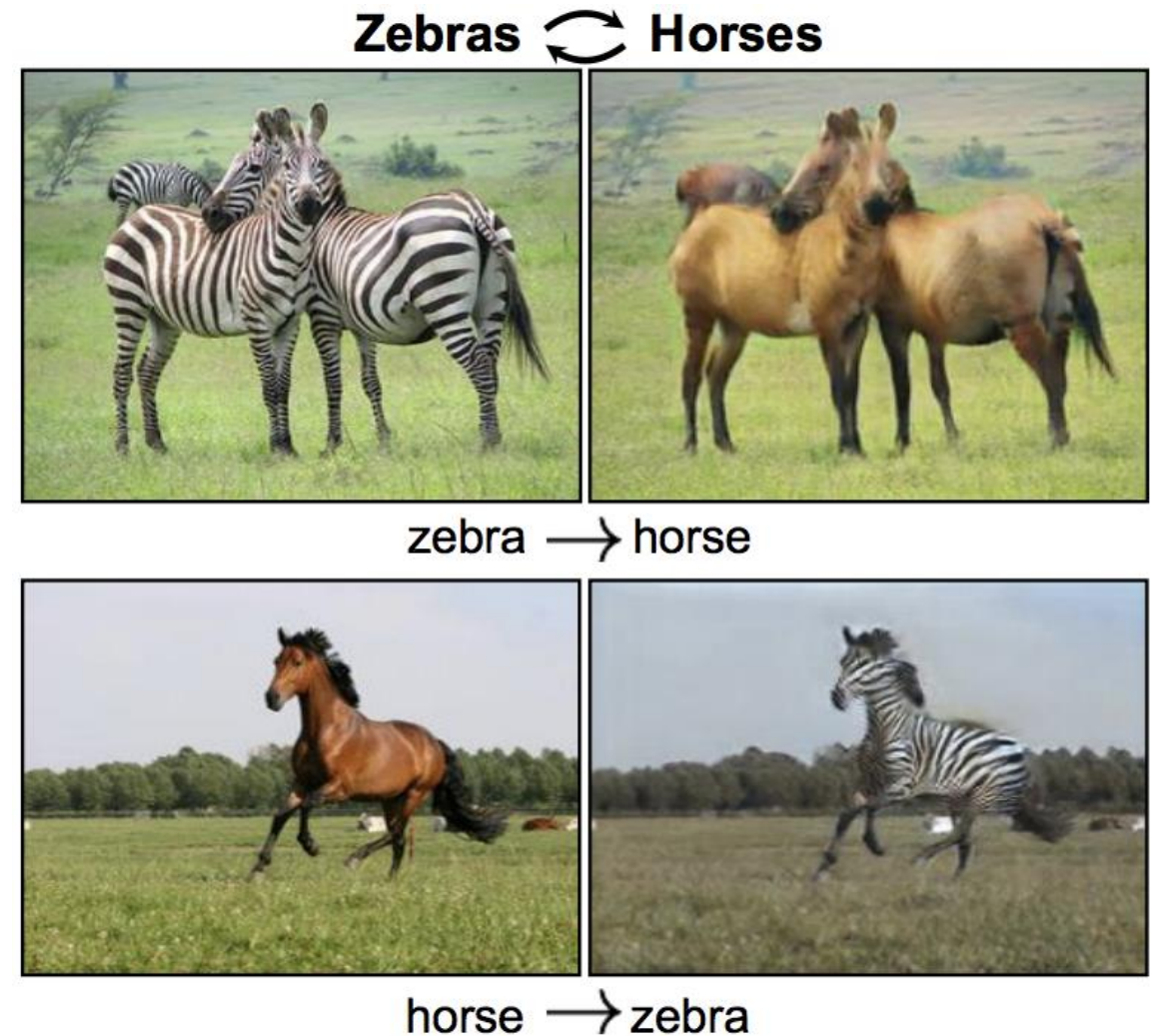


Example of the Results From Different Super-Resolution Techniques.

Taken from “Photo-Realistic Single Image Super-Resolution Using a Generative Adversarial Network”

# 9. Image Synthesis

- It is the task of **generating targeted modifications of existing images** or entirely new images.

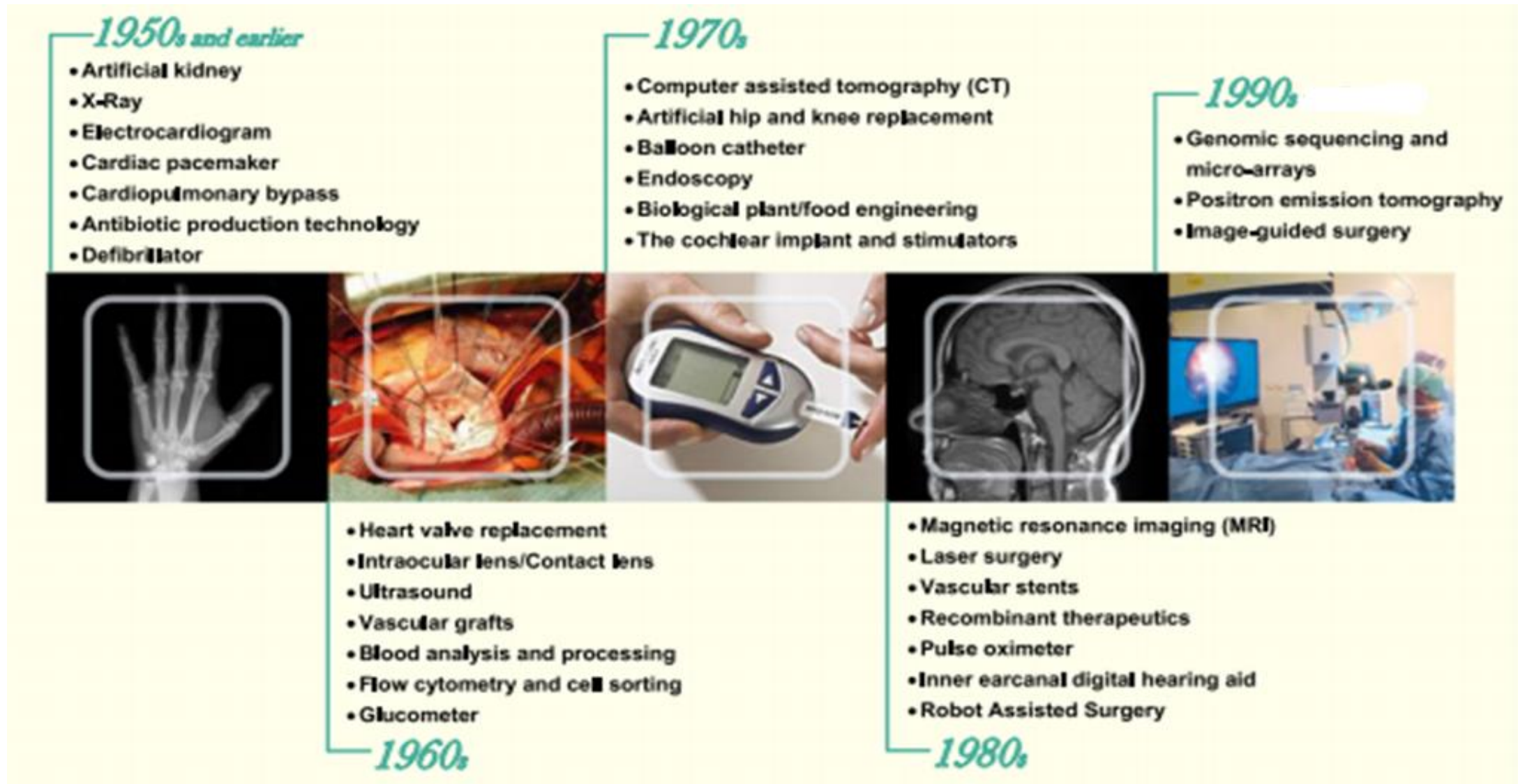


Example of Styling Zebras and Horses.  
Taken from “Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks”



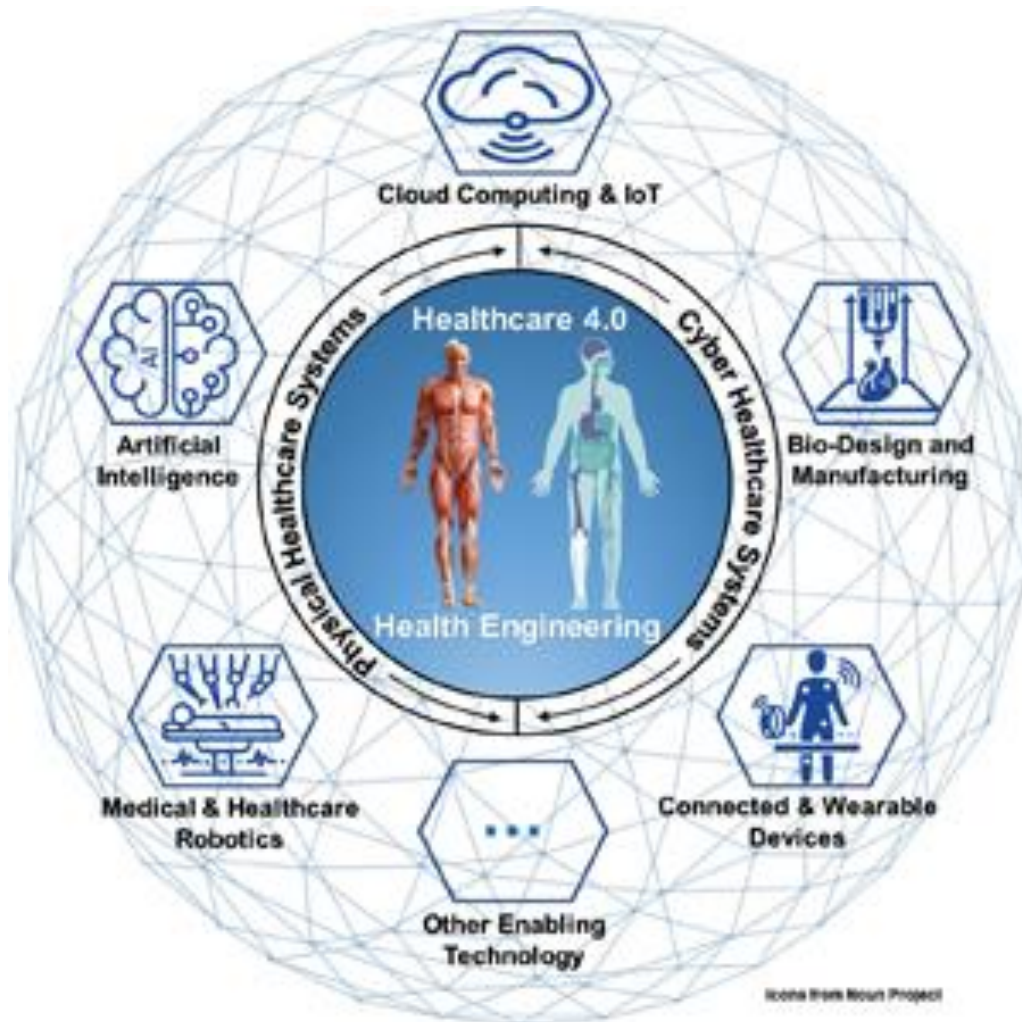
# Medical image (Radiology) and Healthcare 4.0

# Medical and Biomedical Engineering Innovations Across the Decades

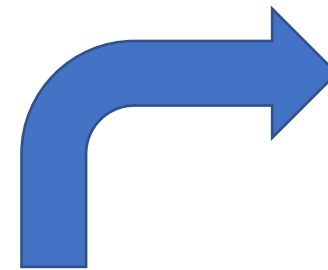


American institute for medical and biological engineering's hall of fame . M. R. Neuman et al., "Advances in medical devices and medical electronics," Proc. IEEE, vol. 100, no. Special Centennial Issue, pp. 1537–1550, May 2012 and Sumber Geng Yang, Introduction to the Special Section: Convergence of Automation Technology, Biomedical Engineering, and Health Informatics Toward the Healthcare 4.0, IEEE Reviews in Biomedical Engineering, January 2018. DOI: 10.1109/RBME.2018.2848518

# The fourth revolution in healthcare technologies (Healthcare 4.0)



As a result, the industry is facing a **second wave of digitalization**, sometimes also referred to as **“Healthcare 4.0”**.



Big Data  
Artificial Intelligent  
Internet of Things  
mobile patient platforms



change every day  
healthcare practice

AI	<ul style="list-style-type: none"> <li>• Device mimics cognitive functions</li> <li>• Since 1950s</li> </ul>
Machine Learning	<ul style="list-style-type: none"> <li>• Algorithms that improve as they are exposed to more data</li> <li>• Since 1980s</li> </ul>
Deep Learning	<ul style="list-style-type: none"> <li>• Artificial neural networks structured in multiple layers to decode imaging raw data</li> <li>• Since 2010s</li> </ul>

# Common medical Imaging



<https://www.radiologyinfo.org>



<http://medicalpicturesinfo.com>



<http://medicalpicturesinfo.com>

## Computed Tomography

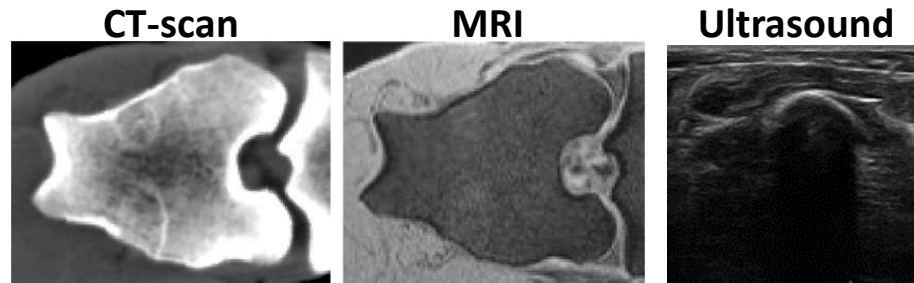
- Radiation exposure
- Expensive
- Limited frequency of use
- Allowed only for patient and cadaver
- Not for healthy human

## Magnetic Resonance Imaging

- Radiation free
- Expensive
- For soft tissues

## Ultrasound

- Radiation free
- Cost effective
- widely available

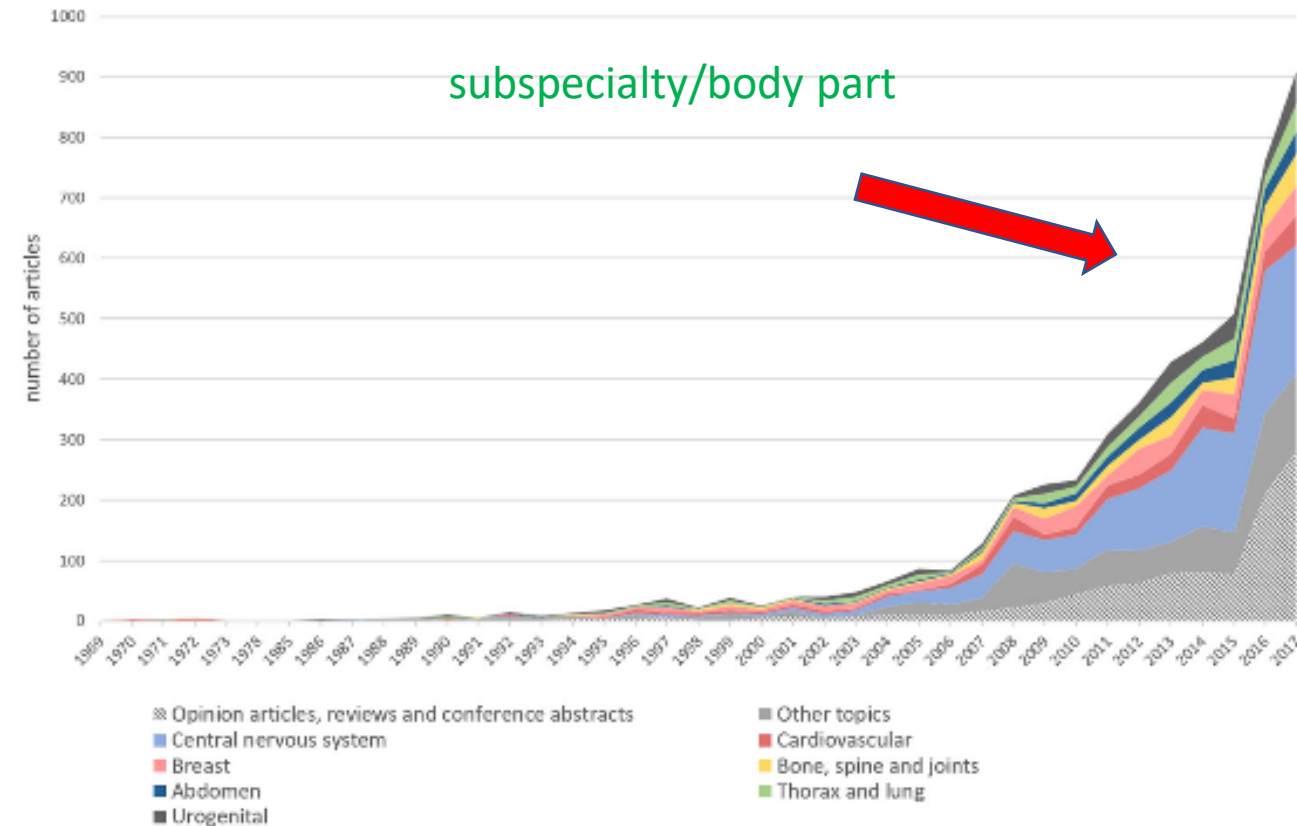
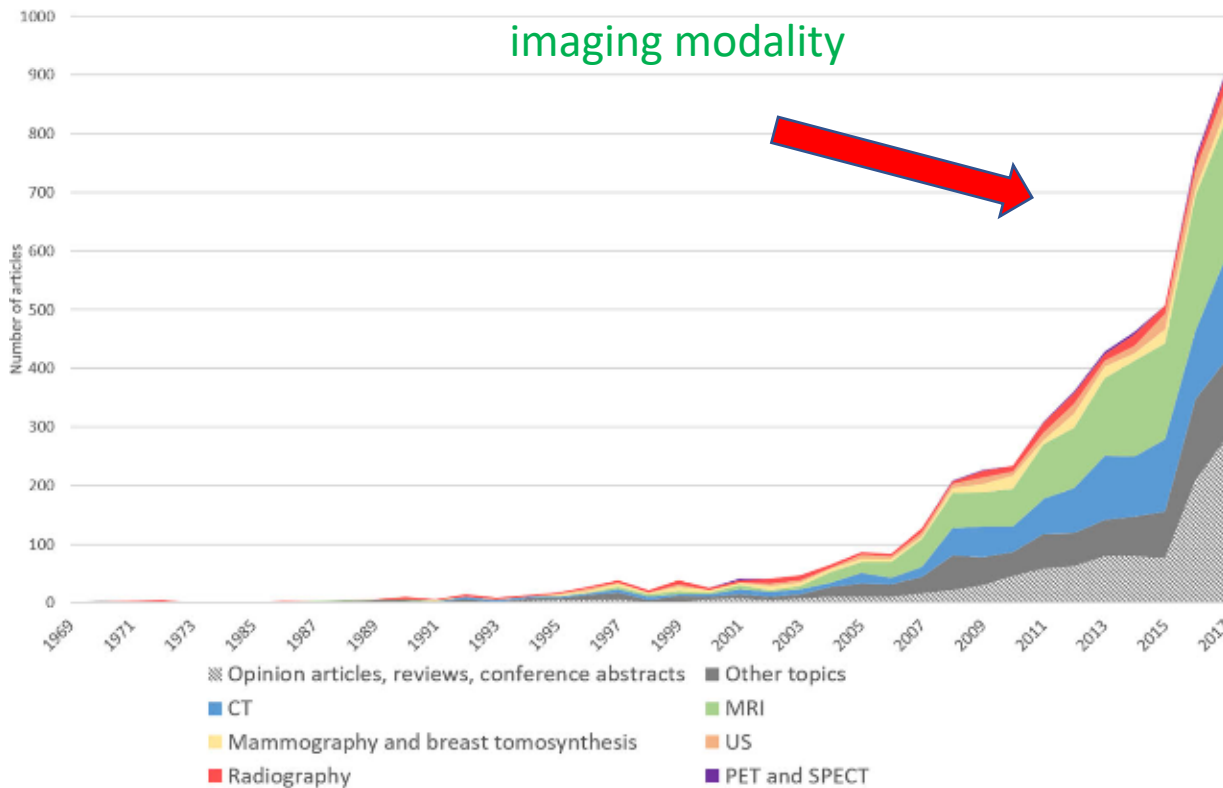


(Rathnayaka et al., 2012)

Information and Data Center, Ministry of Health, Republic of Indonesia (Chalid and Ajeng, 2014)

Medical	Total
MRI	113
Ultrasound	1.441
CT-Scan	410
X-Ray	1.258
Mammography	368

# The great enthusiasm for the development of AI systems in radiology



Over 10 years, publications on **AI in radiology** have **increased** from 100–150 per year to 700–800 per year.



# Artificial intelligence in medical imaging: threat or opportunity?



# Challenges in introducing AI to the radiology department

- *acquiring access and ingesting the data*
- *sufficiently annotating the data*
- the creation of high-quality training datasets
- *storage strategy*
- interpretation of obtained results
- *governance/policy use throughout*
- *legal and ethical*





TERIMAKASIH