

1. Computer Vision - ILSVRC DATASET, RESNET

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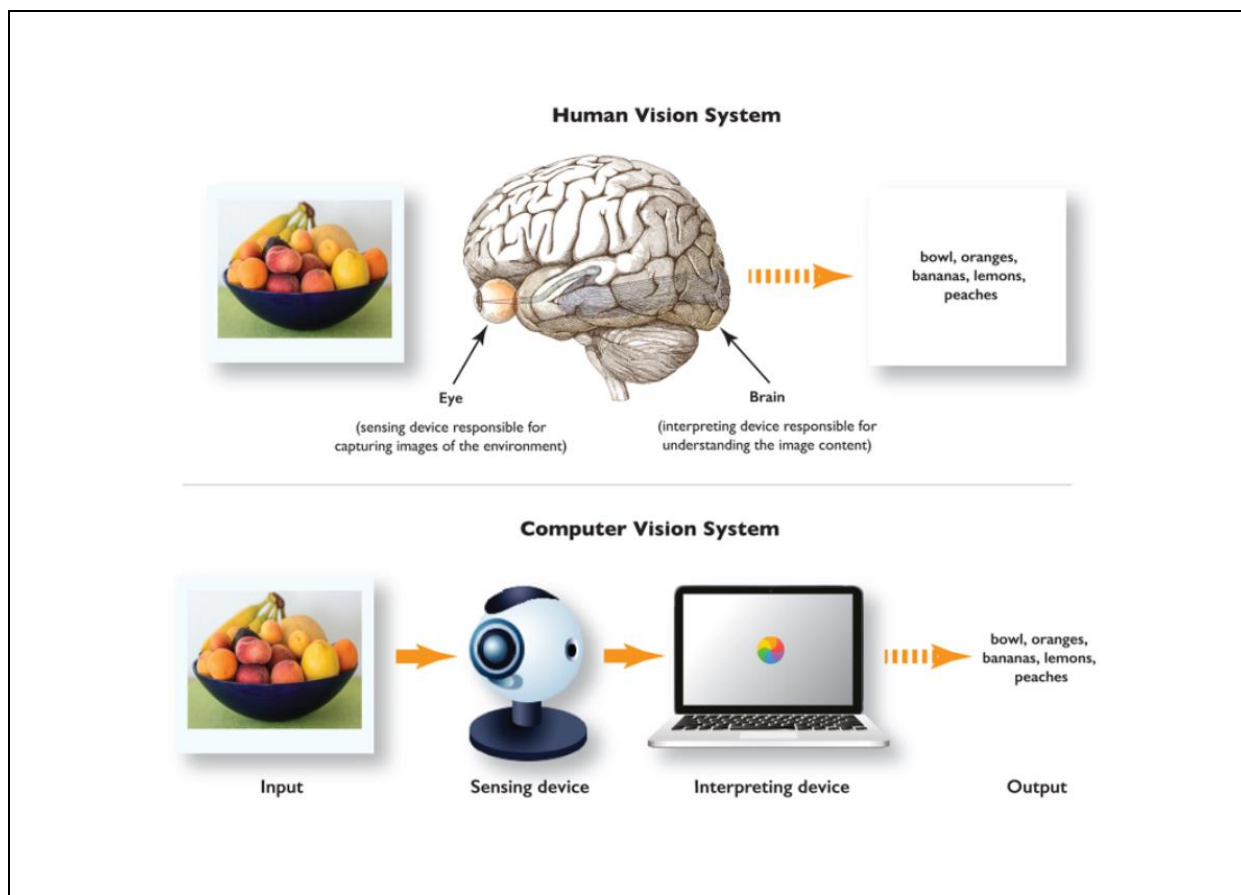
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1. Computer vision

- ✓ Computer Vision is a subfield of Deep Learning and Artificial Intelligence.
- ✓ By using this human can teach computers to see and interpret the world around them.

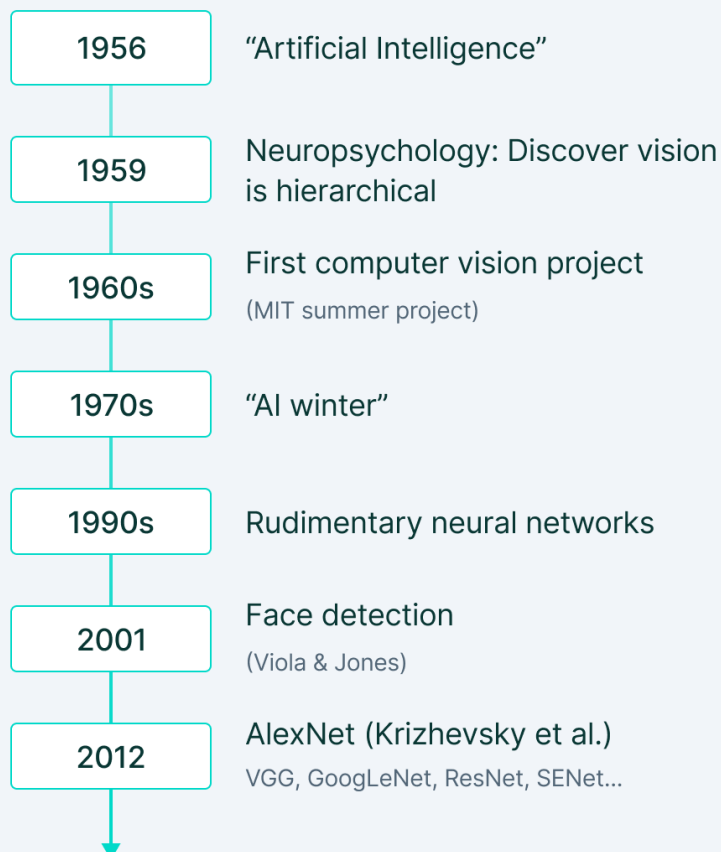
2. Human & computer vision systems



3. History of Computer Vision

- ✓ Machine vision, inspired by animal vision since 1959, evolved with advances in image technology.
- ✓ Key milestones: AI in the 1960s, OCR (Optical Character Recognition) in 1974, and complex topics by the 2000s.
 - Object identification
 - Facial recognition
 - Image Segmentation
 - Image Classification

A brief history of Computer Vision



4. How Computer Vision works?

- ✓ There are mainly three steps,
 - Acquiring an image
 - Processing the image
 - Understanding the image

How does Computer Vision work?



Acquiring an image

Images, even large sets, can be acquired in real-time through video, photos or 3D technology for analysis.



Processing the image

Deep learning models automate much of this process, but the models are often trained by first being fed a thousand of labeled or pre-identified images.

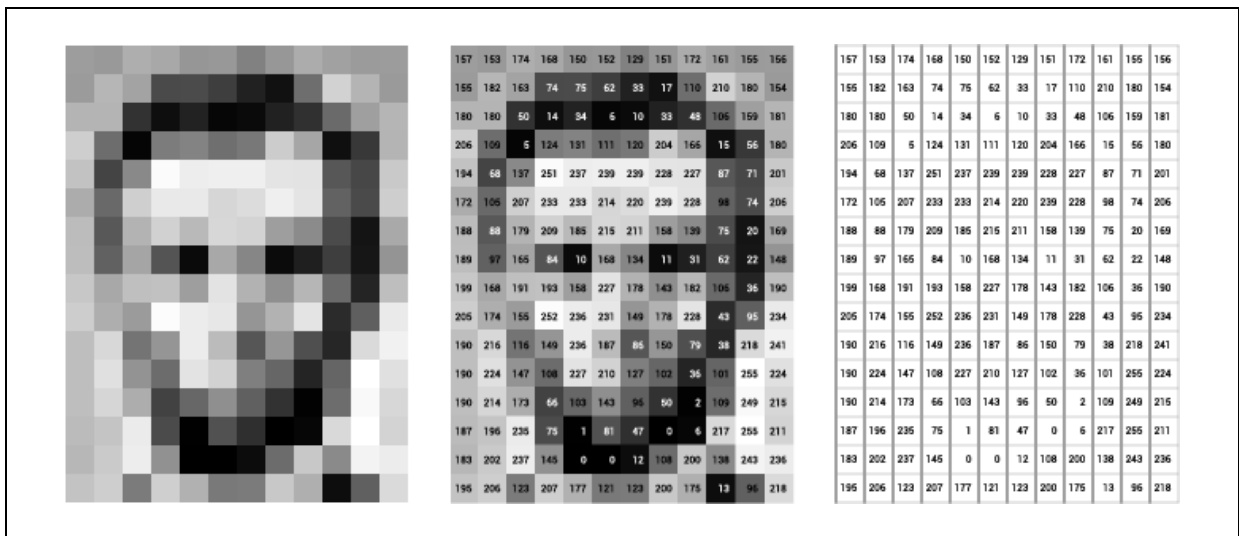
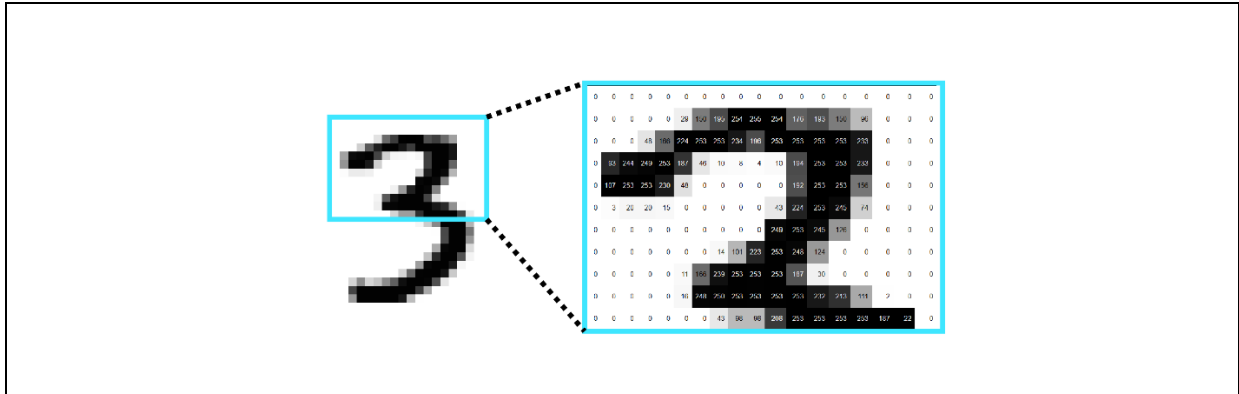


Understanding the image

The final step is the interpretative step, where an object is identified or classified.

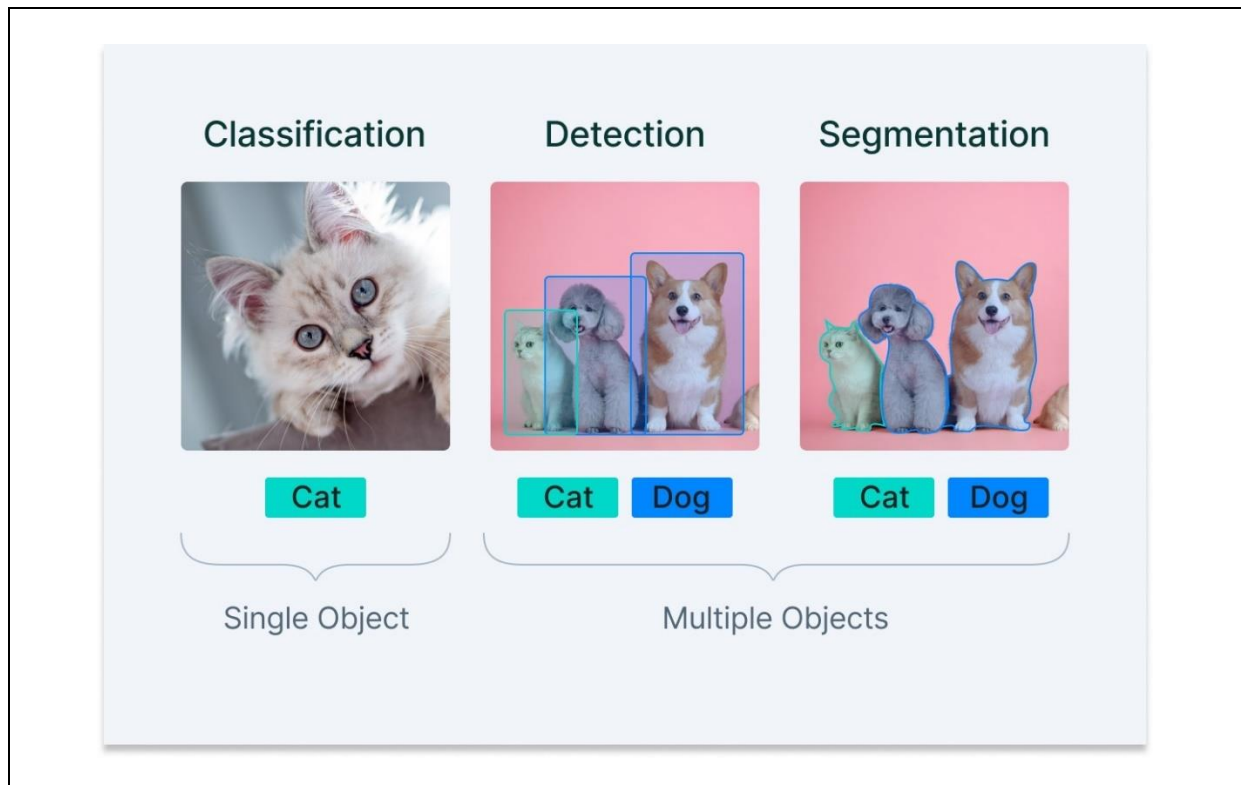
5. Image means pixel to computer

- ✓ An image is group of pixels.
- ✓ Computers process images as pixel arrays.
- ✓ Computer vision uses,
 - Linear algebra for simple tasks and convolutions with pooling for complex tasks.



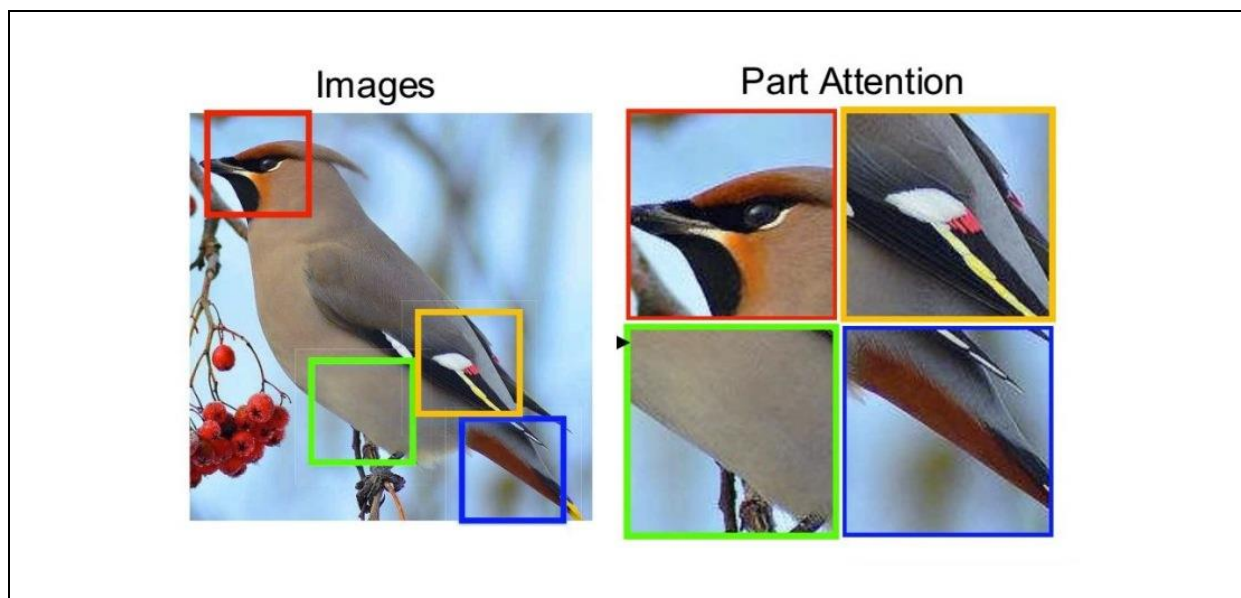
6. Common computer vision tasks

- ✓ Classification
- ✓ Detection
- ✓ Segmentation
- ✓ Face recognition
- ✓ Edge detection & etc



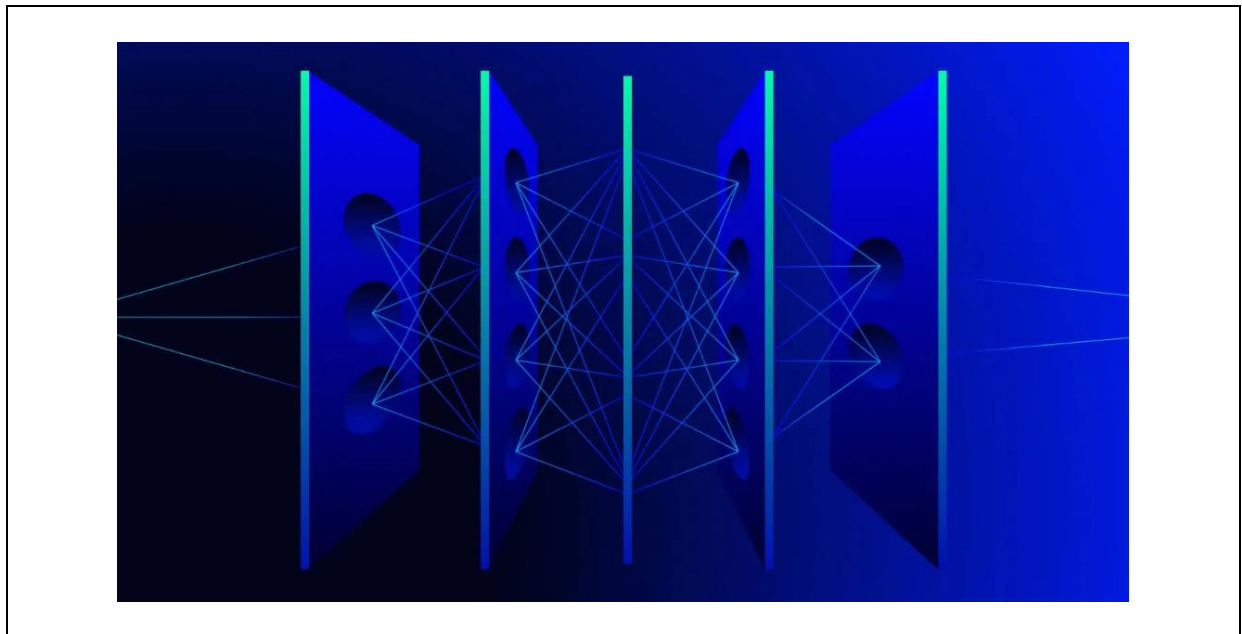
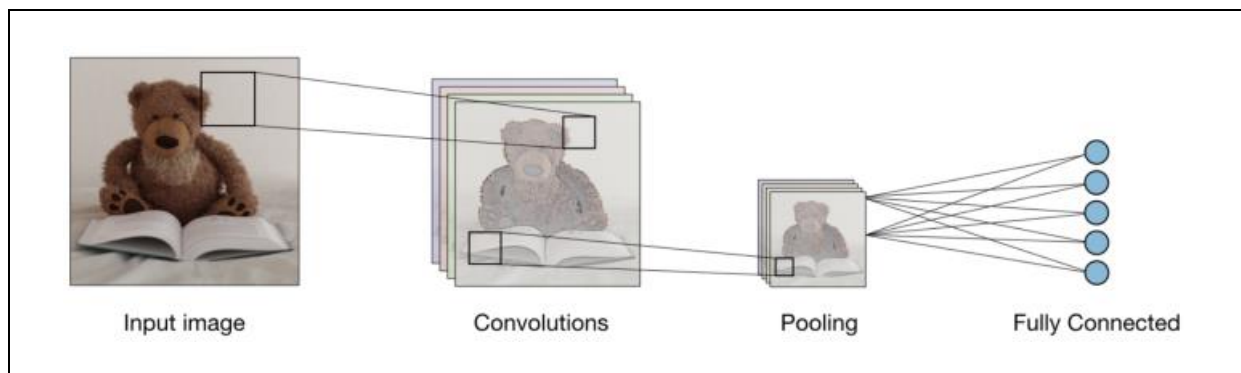
7. Human Image recognition

- ✓ Image recognition means, a computer or machine can see/observe the images.
- ✓ This is something like, **how our eye works**.
 - If we see an image then automatically it **split** into parts of image.
 - In next step our eyes can **analyse** sub-parts individually.
 - By assembling these parts, our eyes **process** and **interpret** the image.
- ✓ The same process was implemented in **CNN** or Convolutional Neural Network.



8. Short Introduction: Convolutional Neural Network

- ✓ CNN is a type of deep learning model.
- ✓ This is commonly used for image recognition and classification tasks.
- ✓ The main parts in CNN,
 - Convolution Layer
 - Pooling Layer
 - Fully-Connected Layer



8.1. Convolutional Layer

- ✓ In this layer, the input image is convolved (In maths: combine one function or series) with a set of learnable filters, also known as kernels.
- ✓ Every filter extracts different features from the input image by performing element-wise multiplication and summation operations.

8.2. Pooling layers

- ✓ Pooling layers reduce the size of the input by taking either the maximum or average value from small regions.
- ✓ This makes the network faster and helps prevent overfitting.

8.3. Fully Connected Layer

- ✓ Connects every neuron in the previous layer to every neuron in the next layer.
- ✓ This layer learns global features.

9. Image recognition task by CNN

- ✓ CNN having layers to recognize the image.
- ✓ Every layer can learn and delivery each task as,
 - The first layer can learn to detect basic edges and shapes.
 - The second layer can recognize these more detailed features.
 - The third layer can recognize parts of objects etc.
- ✓ This learning process helps the network to build a detailed understanding of the data.

10. ImageNet

10.1. The Birth of ImageNet

- ✓ Once upon a time, in the mid-2000s, a brilliant researcher named **Fei-Fei Li**.
 - <https://profiles.stanford.edu/fei-fei-li>
 - https://en.wikipedia.org/wiki/Fei-Fei_Li
- ✓ She had an idea like, just as **humans learn** by seeing countless objects and scenes throughout their lives.
- ✓ Her idea to make **computers see** and understand the world like humans.
- ✓ So, here needed a vast amount of **data**.
- ✓ **She** and her **team** aimed to **create** ImageNet.
- ✓ Started **gathering** images and created dataset
- ✓ The ImageNet is a very large collection of a **dataset**.
 - <https://image-net.org/index.php>

10.2. Dataset Information

- ✓ Gathering images, labelling them accurately was really a challenging task.
- ✓ Amazon Mechanical Turk thousands of team members manually verify and annotate the images.

10.3. Dataset Information

- ✓ This dataset contains,
 - More than **14** million images
 - More than **22** thousand groups or classes.
 - More than **1** million images that have bounding box annotations

10.4. The Birth of the Competition: ILSVRC

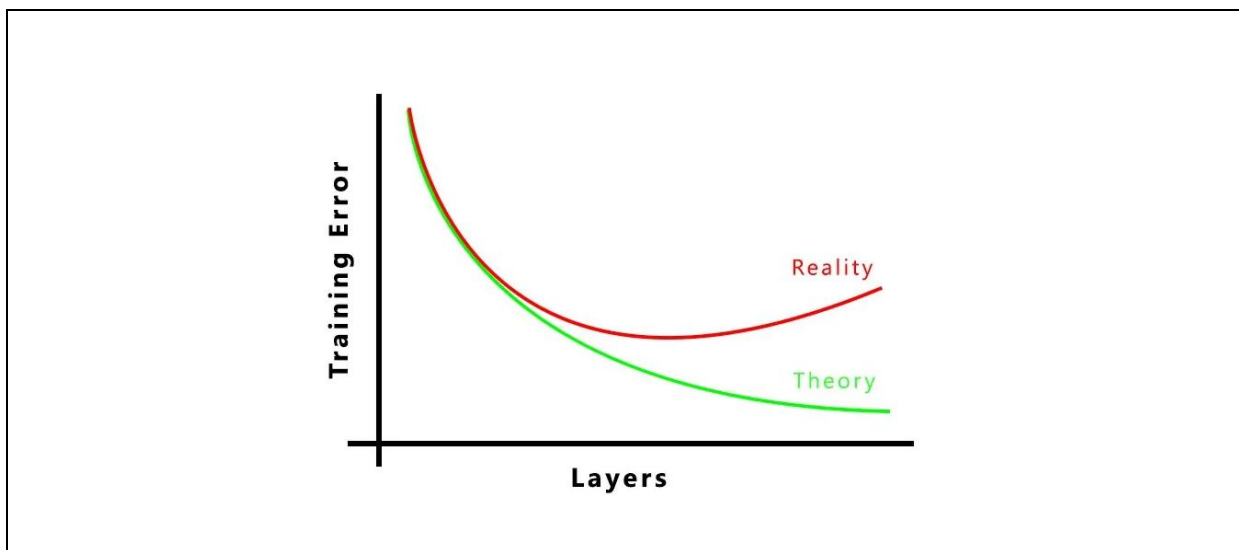
- ✓ ILSVRC the full form is,
 - ImageNet Large Scale Visual Recognition Challenge
- ✓ In 2010, Fei-Fei Li and her team introduced the ImageNet Large Scale Visual Recognition Challenge (ILSVRC).
- ✓ This annual competition invited researchers from around the world to develop algorithms that could classify and detect objects within the ImageNet dataset.
- ✓ It quickly became the benchmark for measuring progress in computer vision.

11. The problem with deep neural networks

- ✓ To get the better results, the architecture used to add more and more layers, but this can lead to vanishing gradient issues.

12. If we increase more layers then...

- ✓ If we increase more layers (deep networks) then we may get the problem of vanishing/exploding gradients.



13. Vanishing / Exploding Gradients

- ✓ If we add more layers to the neural network then one of the problems is like, gradients can become too small (vanishing) or too large (exploding).
- ✓ Due to this Neural Network cannot learn effectively.

14. How to solve Vanishing / Exploding Gradients problem?

- ✓ We can solve this problem by using **ResNet**.

15. History of ResNet

- ✓ ResNet was implemented by Kaiming He and his team, in 2015.
- ✓ This research team achieved top results for object detection and object detection with localization tasks.

15.1. ResNet Paper

- ✓ This research paper written for Deep Residual Learning for Image Recognition.
- ✓ Find the below link for research paper,
 - <https://arxiv.org/abs/1512.03385>

15.2. Winner of ILSVRC 2015

- ✓ ResNet won the champion for ILSVRC 2015 in,
 - Image classification, detection, and localization

16. ResNet Introduction

- ✓ ResNet stands for Residual Network.
- ✓ It's a type of deep learning model.
- ✓ The main goal of ResNet (Residual Network) is to address the vanishing gradient problem.
- ✓ ResNet proves that, it is easy to train very deep neural networks.

17. ResNet Example

Program Name ResNet Example
demo1.py

```
import numpy as np

from keras.preprocessing import image
from keras.applications import ResNet50
from keras.applications.resnet50 import preprocess_input
from keras.applications.resnet50 import decode_predictions

model = ResNet50(weights = 'imagenet')

def load_and_preprocess_image(img_path):
    img = image.load_img(img_path, target_size = (224, 224))
    img_array = image.img_to_array(img)
    img_array = np.expand_dims(img_array, axis = 0)
    img_array = preprocess_input(img_array)

    return img_array

img_path = '1.jpeg'
img_array = load_and_preprocess_image(img_path)
predictions = model.predict(img_array)
decoded_predictions = decode_predictions(predictions, top = 1)[0]

for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")
```

Output

goldfish