

Advanced Computer Vision

Week 01

Sep. 2, 2022
Seokju Lee

Human Visual System



Learning Human Brain from "Baby": The Purest Natural Intelligence



아기성장보고서, EBS

✓ Visual capabilities:

- "What-path": object recognition
- "Where-path": localization

Learning Human Brain from "Baby"



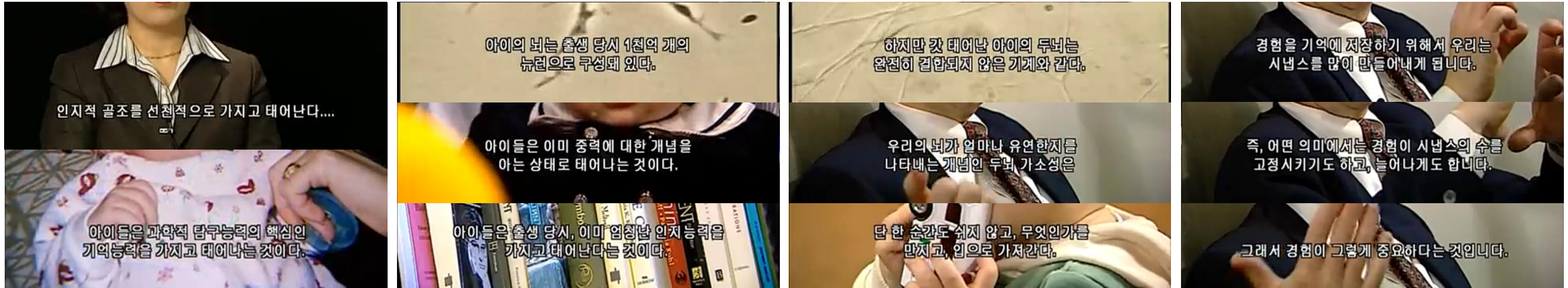
아기성장보고서, EBS

✓ Interaction & Adaptation:

- Environmental interaction with vision, motion, audio, ...
- Ego-centric understanding \Rightarrow Environmental adaptation

Synchronized & co-embedded
multimodal sensing knowledge

Learning Human Brain from "Baby"



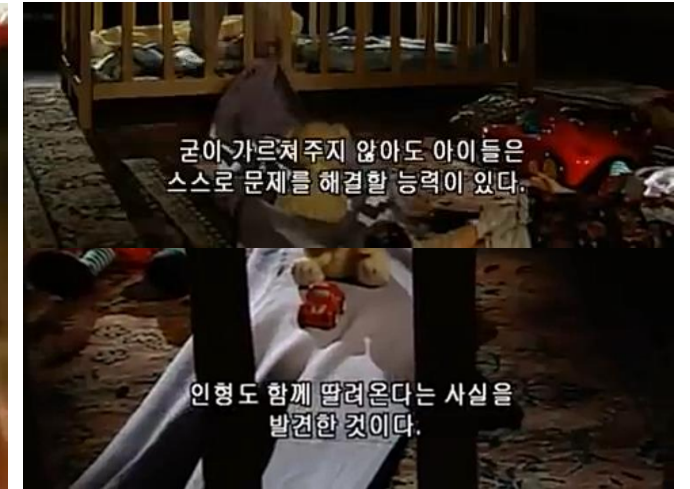
아기성장보고서, EBS

✓ **Powerful memory & Flexible update:**

- Innate memory and well-designed cognitive process
- Life-long abstraction and dynamically-updating synapses from experience

c.f., Pretrained model
& Transfer learning

Learning Human Brain from "Baby"



아기성장보고서, EBS

✓ Reasoning & Learning:

- Reasoning based on uncertain and noisy information
- Learning functional and contextual information

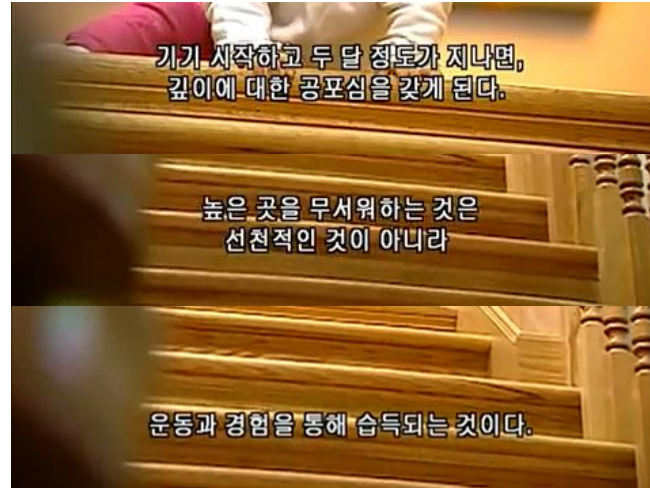
Learning Human Brain from "Baby"



아기성장보고서, EBS

- ✓ **Active & Self-supervised learning:**
 - Curiosity-driven active learning
 - Reward-driven self-supervised learning

Learning Human Brain from "Baby"



아기성장보고서, EBS

✓ We Feel, therefore We Learn:

- "Emotions are inherently linked to and influence cognitive skills such as attention, memory, executive function, decision-making, critical thinking, problem-solving and regulation, all of which play a key role in learning." [1]

Could be one of the most
difficult challenges for AI

[1] M. H. Immordino-Yang, et al., "We Feel, Therefore We Learn: The Relevance of Affective and Social Neuroscience to Education" Mind, Brain, and Education 2007.

How about Robotic Intelligence so far?



Task-Specific Perception

✓ Computer vision:

- Traditional approaches for 3D pose estimation and segmentation

✓ Human operator:

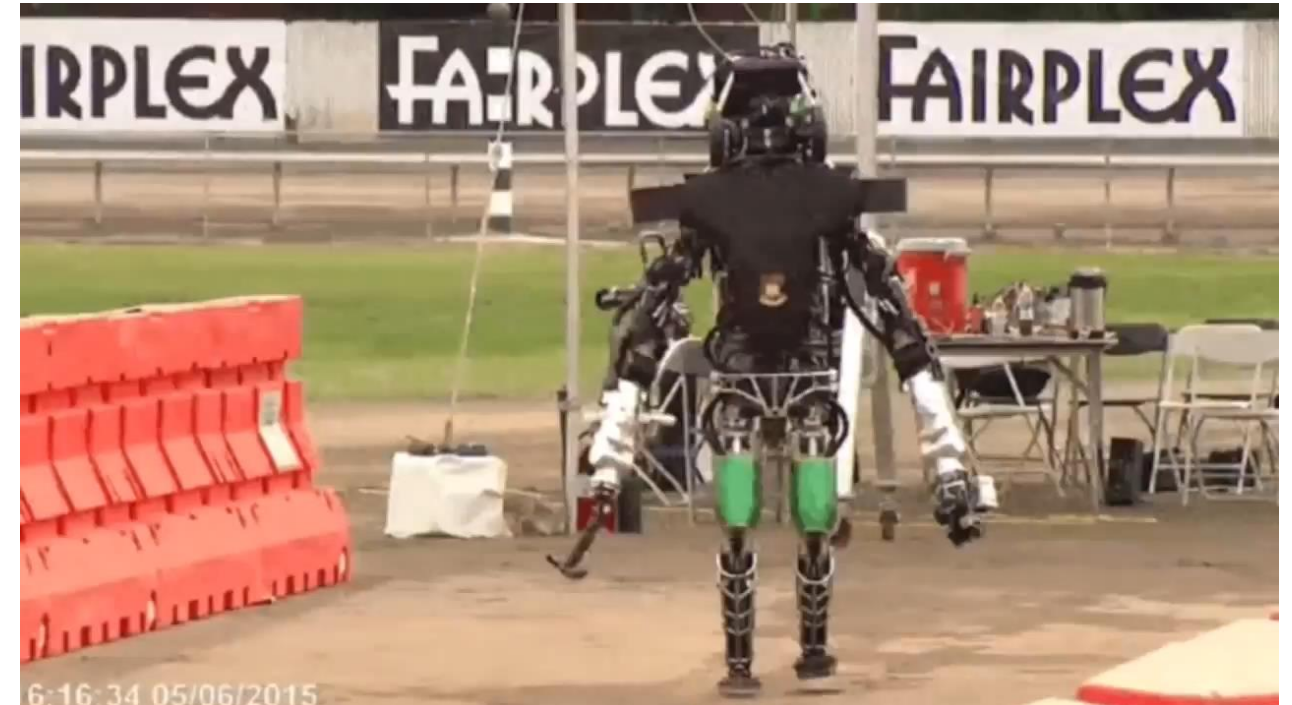
- Planning & Decision making

Lack of Reasoning!

Please focus on the **moment** when the robot **fails!** → **high uncertainty**



KAIST Wins DARPA Robotics Challenge (2015)



Fails compilation

Toward Human-like Intelligence

- ✓ **Reinforcement learning for recognition and manipulation:**
 - Different manipulation strategies for object shapes
- ✓ **Two “Deep Nets” by 800,000 training:**
 - One net for predicting the grasping
 - Another net for estimating the effectiveness of grasping (*meta-learning* approach)

“Learning to learn”

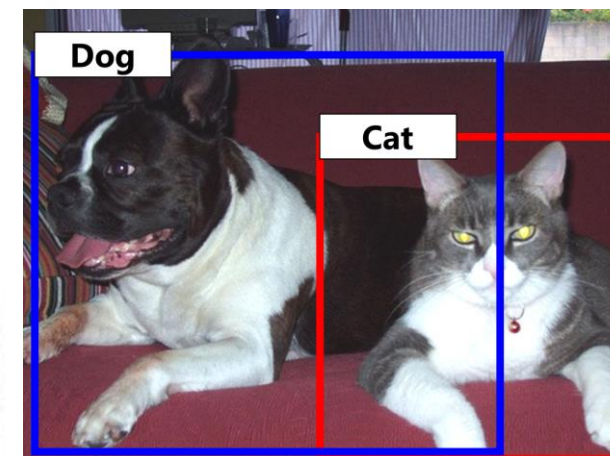
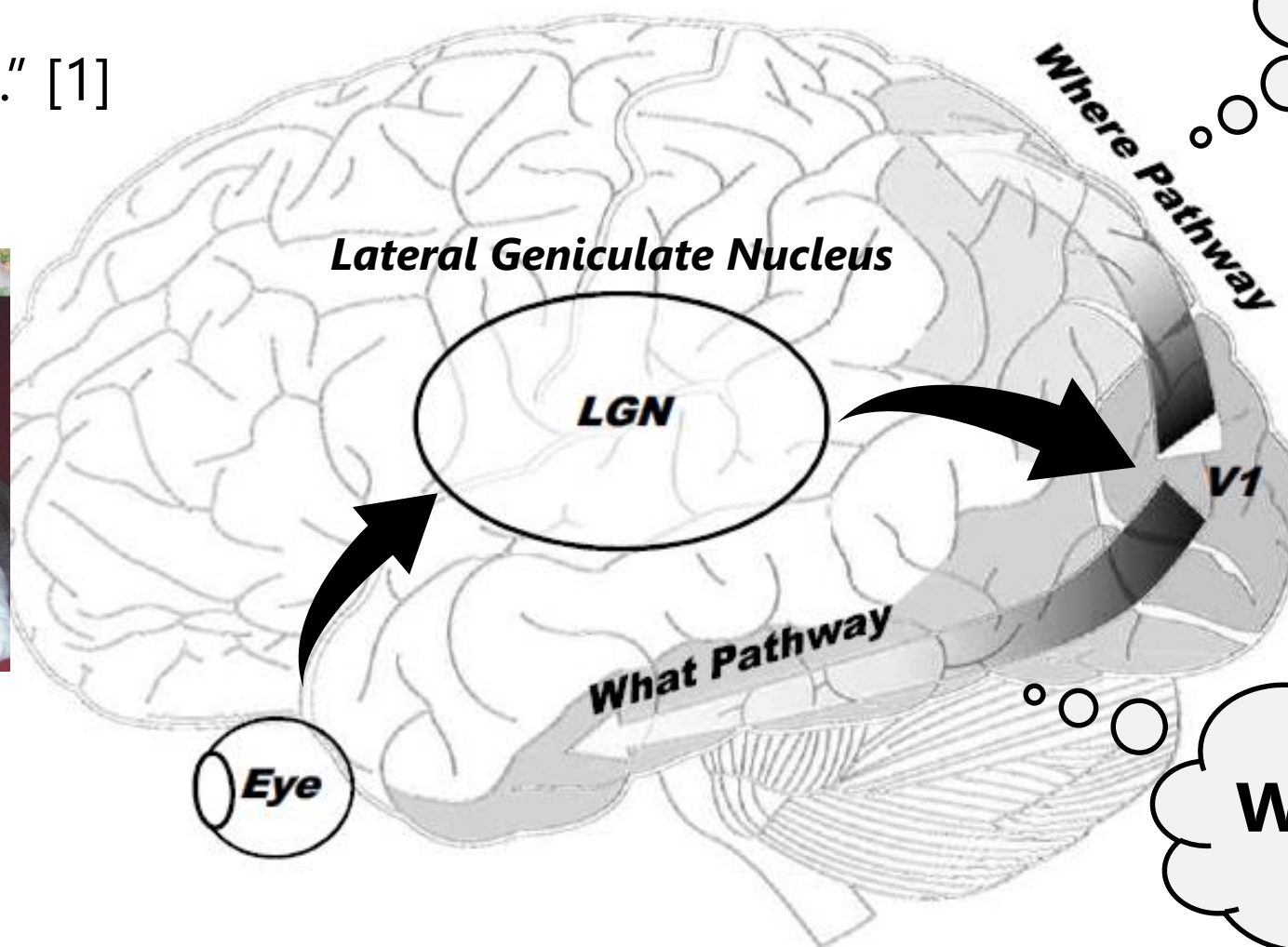


기계와의 대결, KBS (14:10~, [Google AI Blog](#), 2016)

Back to Human Visual System

Human Visual System (HVS)

"More than **half** of neocortex in humans is devoted to **vision**." [1]



[1] Barton, Robert A. "Visual specialization and brain evolution in primates." *Proceedings of the Royal Society of London* (1998).

[2] M. A. Goodale, et al., "Separate visual pathways for perception and action." *Trends in Neurosciences* (1992).

HVS: Very Robust to Any Changes

Scale change



3D viewpoint variation



Illumination change



Environmental change



Image blur, occlusion, clutters, ...



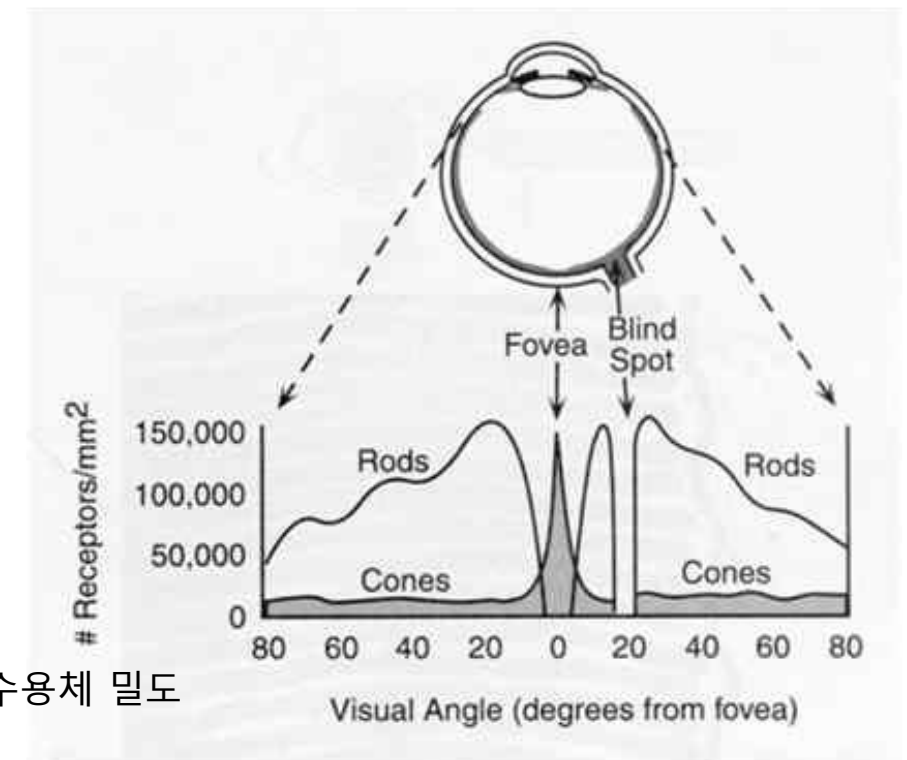
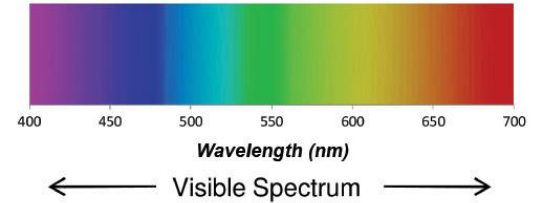
Characteristics of Human Eye

✓ **The human eye is sensitive to only a small set of wavelengths:**

- Rods (간상세포, B/W) and Cones (원추세포, Color)

✓ **The pigments in the cones have peak sensitivities at:**

- **Red** (650nm): L-cone
 - **Green** (530nm): M-cone
 - **Blue** (425nm): S-cone
 - The sensitivities are independent.
- ✓ **The human eye can distinguish about 380,000 colors:**
- 128 Hues (색상, color)
 - 130 Tints (채도, the addition of white, saturation)
 - 16-23 Shades (명도, the addition of black)



→ **Rods are more sensitive to light than cones**

Mechanism of Evolution: Natural Selection

Rods are more **sensitive to light** than cones

→ Easier to **adapt to dark** environments!



뉴스홈 | 최신기사

야생동물 로드킬 5~6월에 빈번..."상향등" 켜지 말아야"

송고시간 | 2017-05-22 14:55



성혜미 기자

기자 페이지

(세종=연합뉴스) 성혜미 기자 = 고속도로와 국도에서 고라니 등 야생동물이 차에 치여 죽는 '로드킬(Road-kill)'은 5~6월에 가장 많은 것으로 나타났다.

동물 발견 시 핸들 급조작, 급브레이크, **상향등**을 사용하면 안 된다.

특히 **상향등**을 사용하면 동물 시력에 장애가 발생해 차량으로 돌진할 위험이 있기 때문에 경적을 울리며 서행해야 한다.

Mechanism of Evolution: Natural Selection

Rods are more **sensitive to light** than cones

→ Easier to **adapt to dark** environments!



Dual camera approach:
One for cones, and the other one for rods

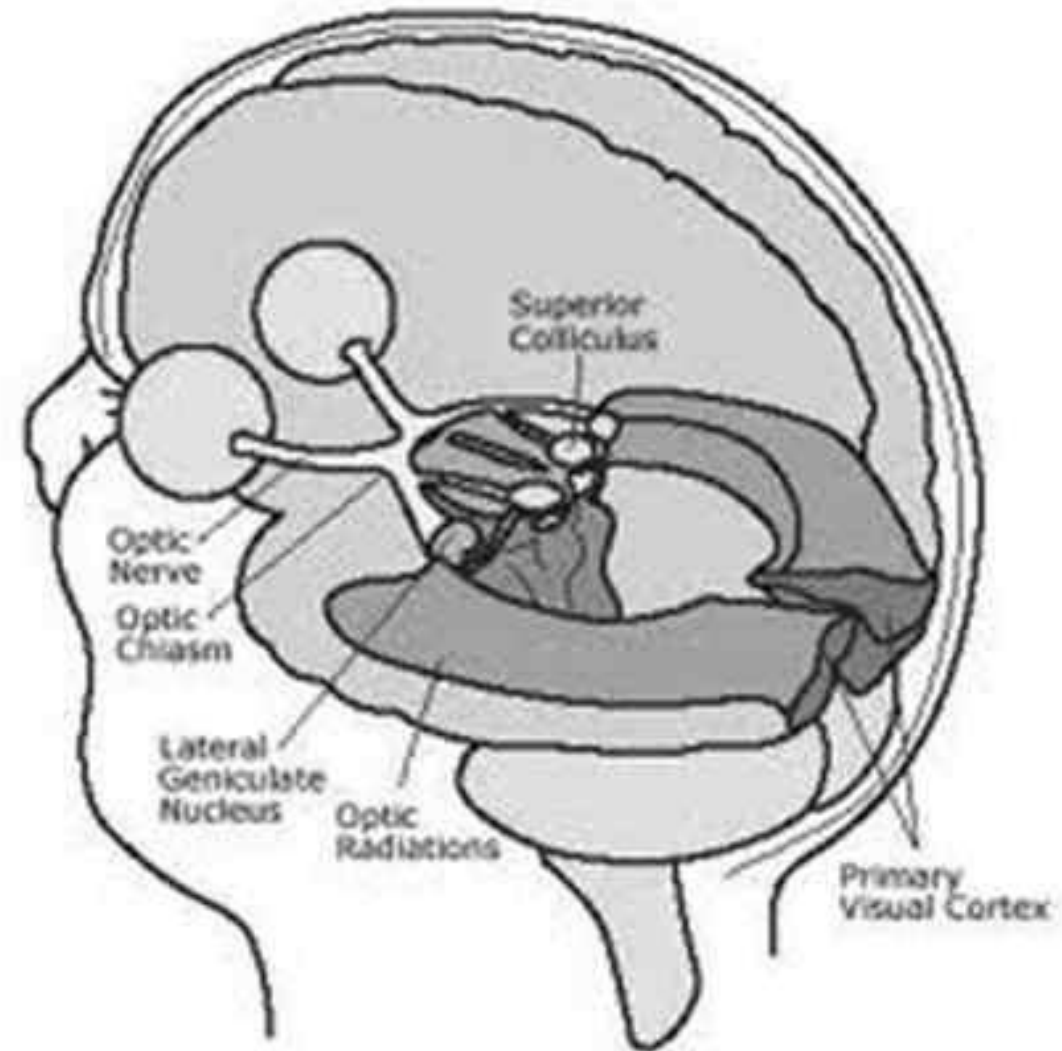


iPhone 11 - Night Mode (2020)

HVS: Where-Path

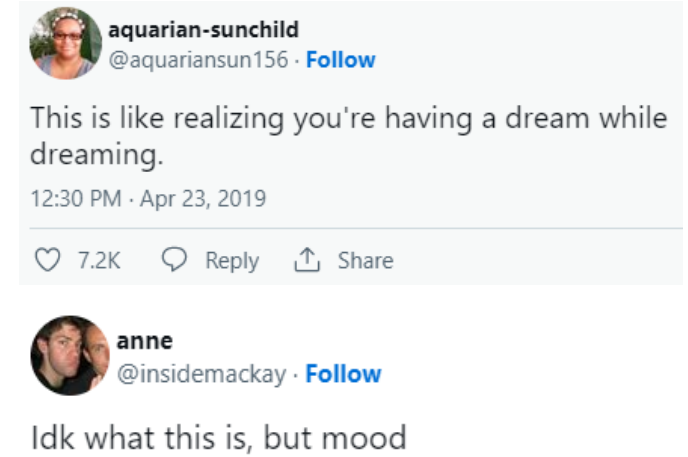
✓ Integration of multiple cues

1. Stereo: depth by two eyes
2. Motion
3. Shading
4. Perspective projection
5. Focus
6. Shadow
7. Position relative to the horizon
8. Relative size
9. Familiar size
10. Texture gradients
11. Aerial perspective
12. Superposition



HVS: What-Path

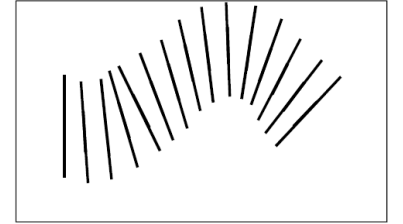
Q. Can you recognize anything in this picture?



→ People can't figure out what is happening in this photo.

HVS: What-Path

- ✓ Object is defined by the **presence** or **absence** of **local** visual properties.

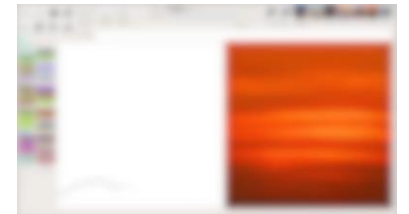


Your HVS constructs all the lines you see and **something more!**

- ✓ Category is described by **object parts** and their **contextual relations**.



AI Draws Images
from Sketch

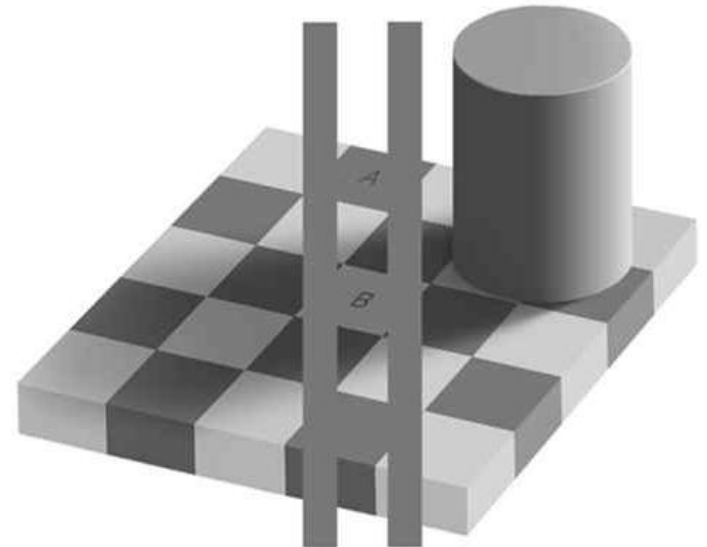
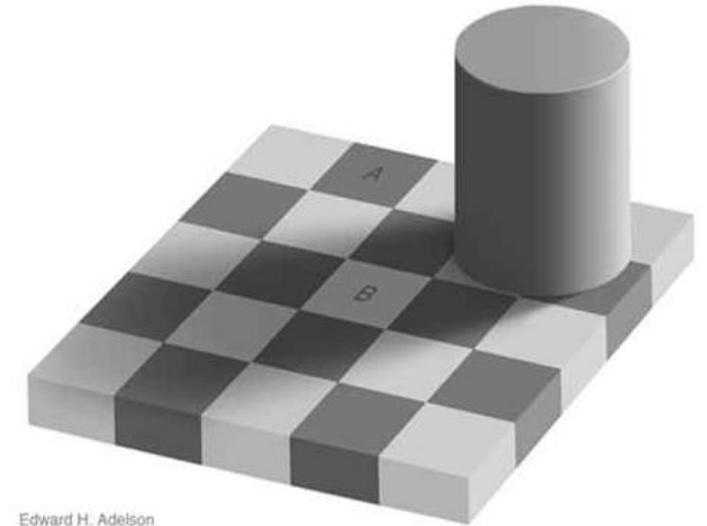


- ✓ Object is linked with **visual attention** and **figure-ground** (형태-배경 조직화).



Human Visual System (HVS): Summary

- **We see effortlessly**
 - 70% of our brains for visual perception
 - Perceiving “the visual world as it is”?
 - A reconstruction of the visual world within our brains
 - Relatively large Field of View (FoV) (200 x 135°)
- **Some limitations of HVS**
 - Subject to illusions
 - Quantitatively imprecise
 - Limited to a narrow range of frequencies of radiation
 - Less than 0.1% of the energy that reaches our eyes.
 - Passive
 - Some active systems: bat acoustic imaging systems



What is "Computer Vision"?



2-D images of the world



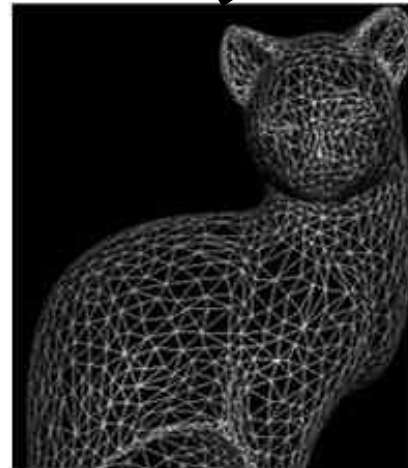
Image synthesis

- **Where**

- Structure and surface properties
 - Geometry
 - Reflectance & roughness
 - Lighting

- **What**

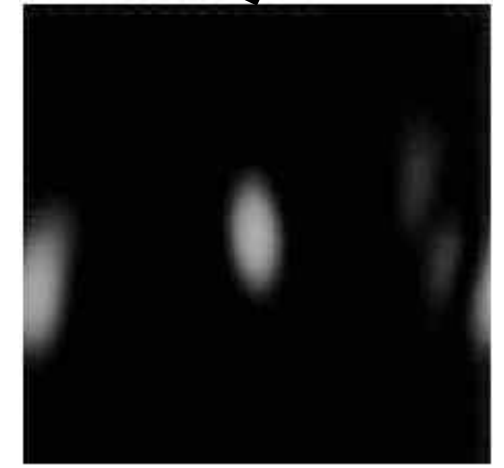
- Object recognition



Geometry



Reflectivity



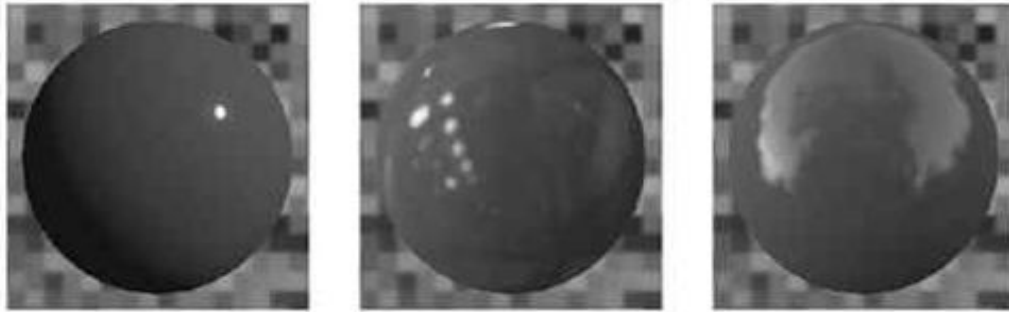
Lighting

Graphics
(rendering)

Why is Computer Vision is Difficult?

✓ Many factors influencing the image intensity:

- Lighting, geometry, reflectivity, sensor, surface roughness, and etc.



• Ambiguities:

- Photometric
- Geometric: 3-D to 2-D projection



• Intractability:

- Huge amounts of data
- High computational complexity

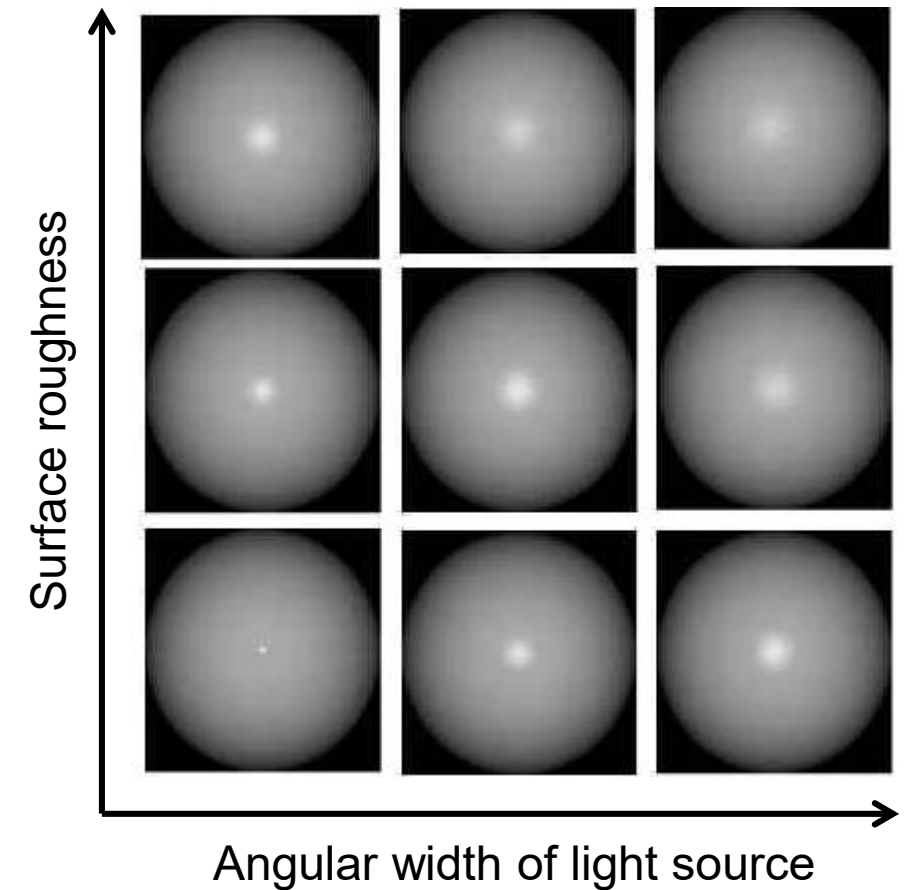


Image Processing with Python

Python Functions

How to define functions?

→ Plot image (show_image) & close the plot (close_image)

```
def show_image(i, img):  
    plt.figure(i)  
    plt.imshow(img)  
    plt.xticks([]); plt.yticks([])  
    plt.ion(); plt.show()  
  
def close_image(i):  
    if i == 0:  
        plt.close('all')  
    else:  
        plt.close(i)
```

```
def do_sum(a, b):  
    output = a + b  
    return output  
  
def do_subtract(a, b):  
    output = a - b  
    return output  
  
def do_multiply(a, b):  
    output = a * b  
    return output
```

Basic Python Library: NumPy

Operations

→ There are many strong operation functions for “**multi-dimensional arrays**”

```
import numpy as np

a = np.array([1,2,3])
print(a)          # [2 3 4]
print(a.dtype)    # int64
b = np.array([1.2, 3.5, 5.1])
print(b.dtype)    # float64
print(a**2)
print(a.sum())
print(a.mean())
print(a.min())
print(a.max())
```

Please try basic math operations.

```
a = np.arange(8)
print(a)          # [0 1 2 3 4 5 6 7]
b = a.reshape(2, 4)
print(b)
c = a.T
print(b)

print(a.shape)
print(b.shape)
print(c.shape)
print(b.sum(axis=0))
print(b.sum(axis=1))
```

→ Matrix operation, linear algebra

Please discuss each print line.

Basic Python Library: NumPy

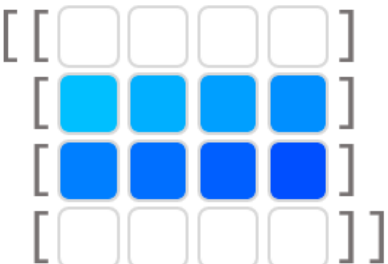
Array slicing

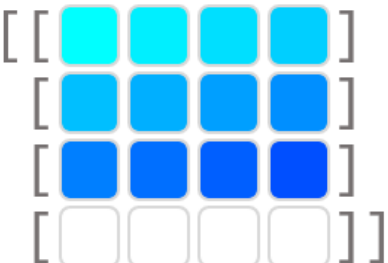
→ Along the first axis

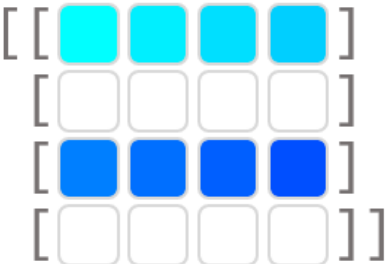
```
import numpy as np

arr = np.array([[1, 2, 3, 4],
                [5, 6, 7, 8],
                [9, 10, 11, 12],
                [13, 14, 15, 16]])

print(arr)
print(arr[1:3])
print(arr[:-1])
print(arr[:,2])
```

`arr[1:3]` = 

`arr[:-1]` = 

`arr[:,2]` = 

Basic Python Library: NumPy

Array slicing

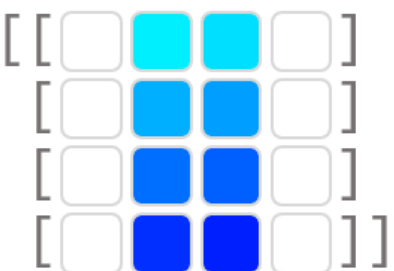
→ Along the second axis

```
import numpy as np

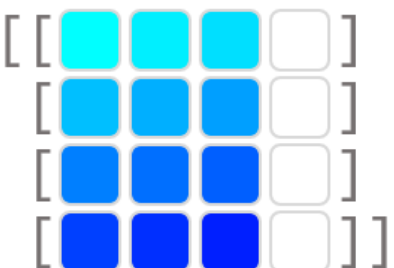
arr = np.array([[1, 2, 3, 4],
                [5, 6, 7, 8],
                [9, 10, 11, 12],
                [13, 14, 15, 16]])

print(arr)
print(arr[:, 1:3])
print(arr[:, :-1])
print(arr[:, ::2])
```

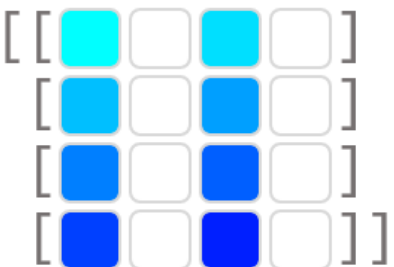
`arr[:, 1:3]` =



`arr[:, :-1]` =



`arr[:, ::2]` =



Basic Python Library: NumPy

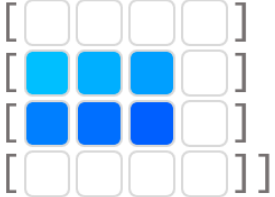
Array slicing

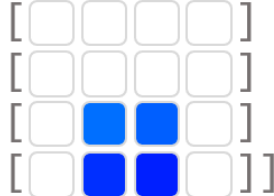
→ Along the first and second axis

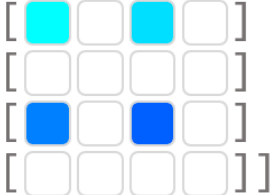
```
import numpy as np


arr = np.array([[1, 2, 3, 4],
                [5, 6, 7, 8],
                [9, 10, 11, 12],
                [13, 14, 15, 16]])

print(arr)
print(arr[1:3, :-1])
print(arr[2:, 1:3])
print(arr[::2, ::2])
print(arr[1::2, 1::2])
```

`arr[1:3, :-1]` = 

`arr[2:, 1:3]` = 

`arr[::2, ::2]` = 

`arr[1::2, 1::2]` = 

Basic Python Library: NumPy

Changing specific values

→ Can be applied with a condition

```
import numpy as np
```

```
a = np.array([[1, 2], [3, 1]])
```

```
b = np.where(a == 1, 10, a)
```

```
print(a)
```

```
print(b)
```

```
import numpy as np
```

```
a = np.array([[0.5, 1.2, 0.9], [1.1, 0.8, 1.4]])
```

```
b = np.where(a < 1.0, 0.0, a)
```

```
print(a)
```

```
print(b)
```

Basic Python Library: NumPy

Concatenate array

→ Along different axis

```
import numpy as np

a = np.array([1, 2, 3])
b = np.array([4, 5, 6])
c = np.array([7, 8, 9])

ab = np.concatenate((a, b))
abc = np.concatenate((a, b, c))

print(ab)
print(abc)
```

```
import numpy as np

a = np.array([[1, 2], [3, 4]])
b = np.array([[5, 6], [7, 8]])

ab_0 = np.concatenate((a, b), axis=0)    # Default
ab_1 = np.concatenate((a, b), axis=1)

print(ab_0)
print(ab_1)
```

Basic Python Library: NumPy

Stack array

→ Along different axis

```
import numpy as np
```

```
a = np.array([[1, 2], [3, 4]])  
b = np.array([[5, 6], [7, 8]])  
c = np.hstack([a, b])  
d = np.vstack([a, b])
```

```
print(a)  
print(b)  
print(c)  
print(d)
```

```
import numpy as np
```

```
a = np.array([[1, 2], [3, 4]])  
b = np.array([[5, 6], [7, 8]])  
c = np.stack([a, b], axis=0)  
d = np.stack([a, b], axis=1)
```

```
print(a)  
print(b)  
print(c)  
print(d)
```

Difference between
concatenate and stack?

Basic Python Library: NumPy

Broadcasting

→ Flexible operations

```
import numpy as np

array1 = np.array([1, 2, 3, 4]).reshape(2, 2)
array2 = np.array([1.5, 2.5])

add = array1 + array2

print(add)
```

What are the rules to enable broadcasting?

Basic Python Library: NumPy

Append array

→ Along different axis

```
import numpy as np

arr = np.array([[ [1, 1], [2, 2] ],
                [ [3, 3], [4, 4] ]])
item = np.array([ [5, 5], [6, 6] ])

print(arr.shape)
print(item.shape)

append = np.append(arr, item.reshape(1, 2, 2), axis=0)

print(append)
```

Image Processing Puzzle



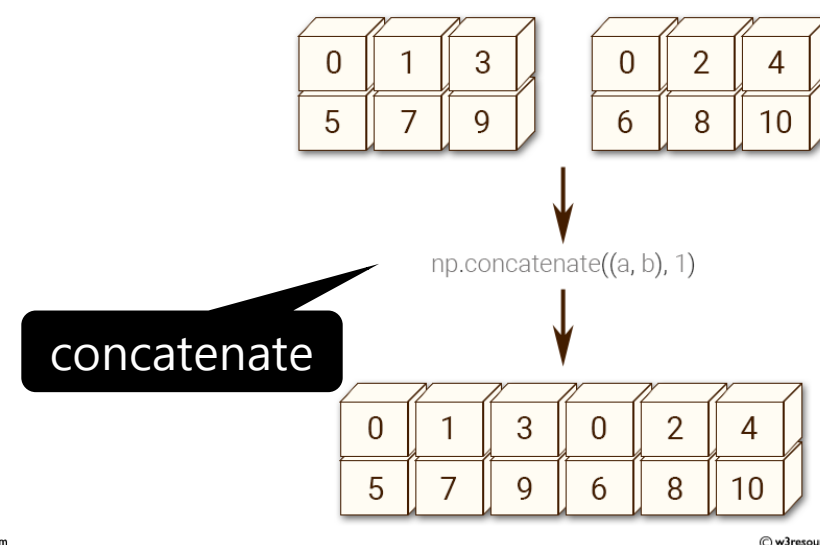
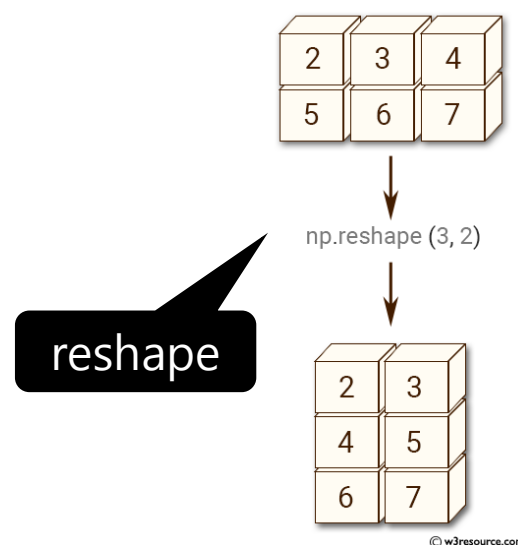
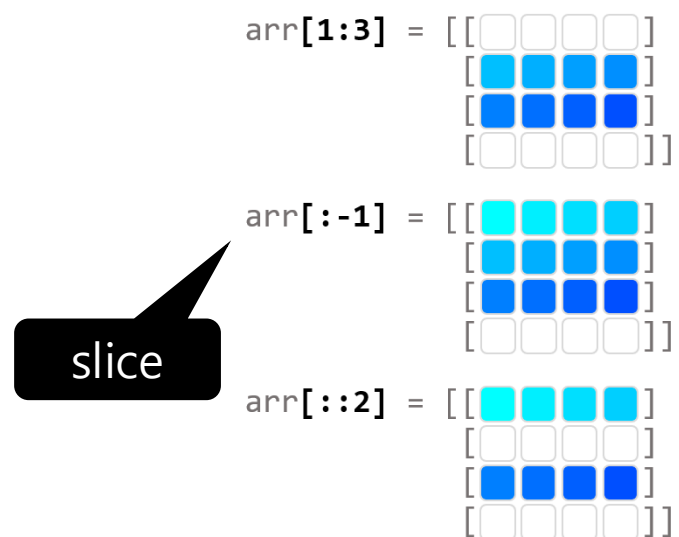
Summary of Previous Lesson

Basics of Python functions

→ How to define and use

Basics of NumPy arrays

- Many useful operations for multi-dimensional arrays
- Please always check the current shape of arrays using the **shape** method!
- Basic math operations: add, subtract, min, max, mean, etc. along different axis
- Basic transformations: slicing, reshape, concatenate, stack, append, etc.



[1] w3resource.com

Now, Let's Play with Images!

Solve image processing puzzles.

→ Multiple image processing missions to make specific images.

→ If you need, please refer below pages.

Numpy: <https://numpy.org/doc/stable/reference/>

Scikit-learn image: <https://scikit-image.org/>

PIL image: <https://pillow.readthedocs.io/en/stable/reference/Image.html>

Codes are available at:

<https://view.kentech.ac.kr/a6fac1a2-7304-409e-85f0-78b1d527034f>

Installed packages

scikit-learn 1.0.2

4 dependencies

numpy

scipy

joblib

threadpoolctl

matplotlib 3.5.1

scikit-image 0.19.2

8 dependencies

numpy

scipy

networkx

pillow

imageio

tifffile

PyWavelets

packaging

numpy 1.22.3

Image Processing Puzzle 1

Now we will try visualization

→ Enables image debugging

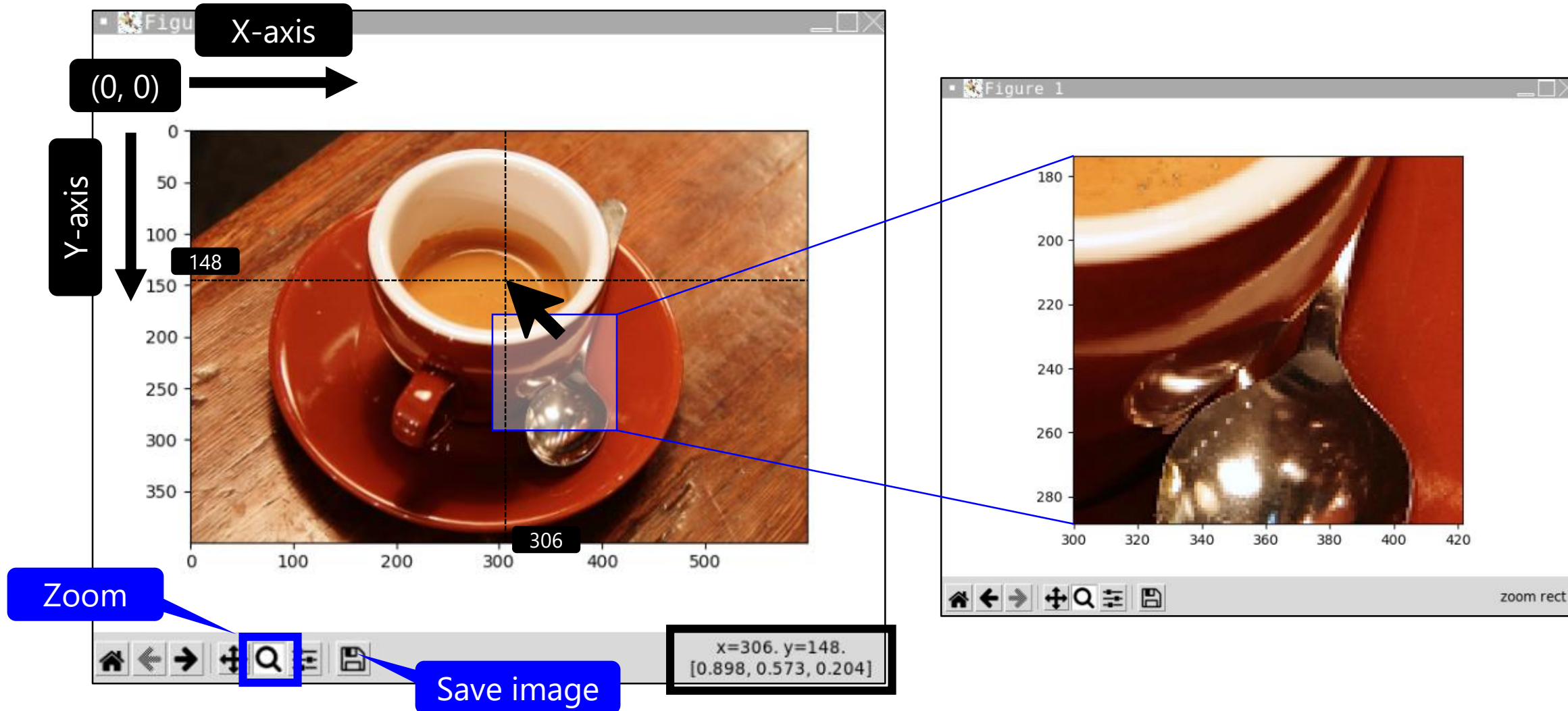


Image Processing Puzzle 1

Discussion 2: shifting image over x-axis

→ Use loop to generate the target image

```
hh, ww, ch = im.shape
```

Check the shape

```
for i in range(ww):
```

Search space: maximum width

```
    print('index:', i)
```

```
    im1 = np.zeros(im.shape)
```

Initialize

```
    im1[:, i:, :] = im[:, :ww-i, :]
```

Assign

```
    if (im1 == gt1).all():
```

```
        break;
```

Break the loop if the target is matched

```
check_same(gt1, im1)
```

Check if the output is correct

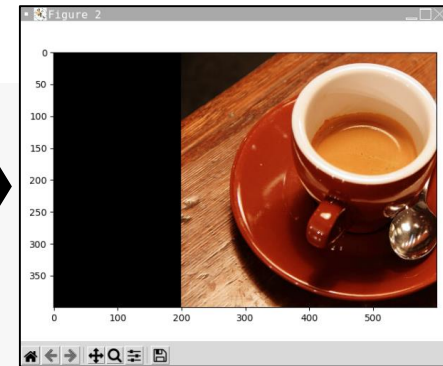
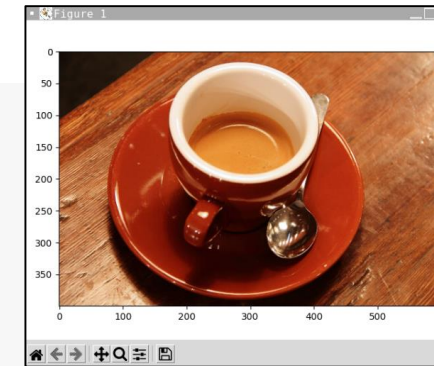


Image Processing Puzzle 1

Discussion 2-2: shifting over both x- and y-axis

→ Use loop to generate the target image → Too slow!

```
flag = False
hh, ww, ch = im.shape
for i in range(hh):
    for j in range(ww):
        print('index:', i, j)
        im1 = np.zeros(im.shape)
        im1[i:, j:, :] = im[:hh-i, :ww-j, :]
        error = np.abs(im1 - gt1).mean()
        if error == 0:
            flag = True
            break;
    if flag:
        break;
check_same(gt1, im1)
```

Search space: both maximum height + width
Use "**nested loop**"

You can check the processing time using the below code lines:

```
import time
start = time.time()
...
print(time.time() - start)
```


Image Processing Puzzle 1

Discussion 2-2: shifting over both x- and y-axis

→ Dimension reduction. Please discuss this code in your report!

```
hh, ww, ch = im.shape
```

```
gt1_h = gt1.sum(axis=(0,2))
```

```
gt1_v = gt1.sum(axis=(1,2))
```

```
jj = (gt1_h == 0).sum()
```

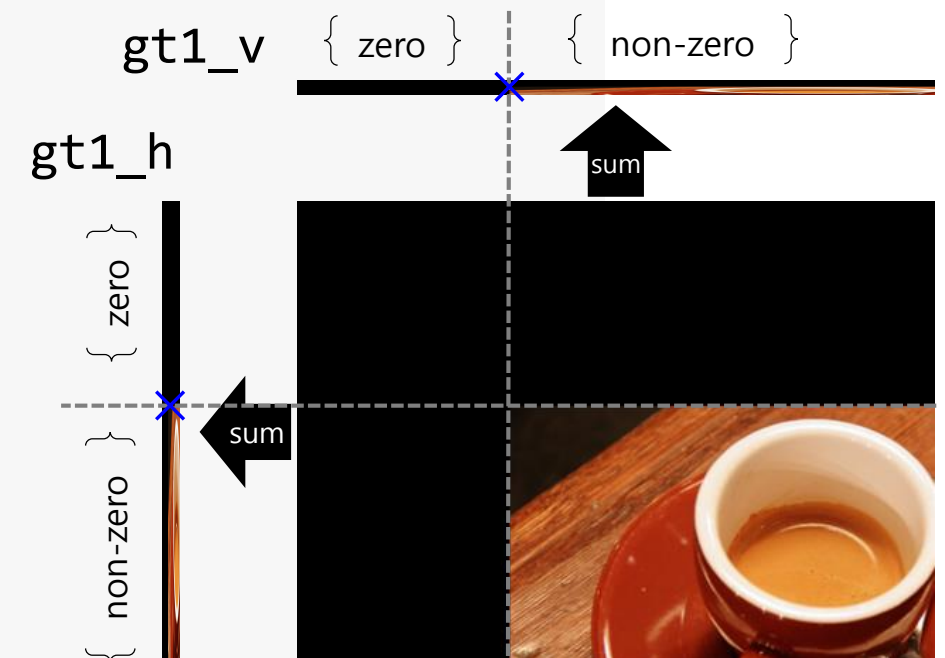
```
ii = (gt1_v == 0).sum()
```

```
im1[ii:, jj:, :] = im[:hh-ii, :ww-jj, :]
```

```
check_same(gt1, im1)
```

Dimension reduction

Automatic binary sum
(= count the number of zeros)



→ Once you are familiar with the **“for-loop”** statement,

Try to implement your code **efficiently** through **dimension reduction**!

Image Processing Puzzle 2

Discussion 4: Image multiplication

→ Multiply a binary mask to black out unnecessary parts.

Discussion 5: Image cropping and resizing

→ Crop a specific region with slicing, and resize the image.

Discussion 6: Photoshop - bilateral filter

→ Let's see how to do "photoshop" with a basic filtering algorithm.

Randomly generated face image (AI model) from
<https://generated.photos/face-generator/new>



Discussion

<Image Processing Puzzle 2>

Please paste your result images (screenshot or save) in this report if needed.

Discussion 2-3: Shifting image

2.7. Try yourself! Please try to use “dimension reduction”.

2.8. Please visualize your output image.

Discussion 4 - Image multiplication

4.1. Please load a ground truth (GT) array from 'samples/puzzle2/gt1.npy'

4.2. Please convert 'im' to look like 'gt1' by multiplying the binary mask.

4.3. What is broadcasting?

4.4. What are the rules for allowing broadcasting?

Discussion 5: Image cropping and resizing

5.1. Crop the image (cat's eye) by array slicing.

5.2. Increase the size of the "cat's eye" image to 400 x 400. Please discuss the difference between "rescale" and "resize".

Discussion 6: Photoshop - bilateral filter

6.1. Please capture random faces from "https://generated.photos/face-generator/new"

6.2. Please apply a "bilateral filter" on the image.

6.3. Please analyze the effect of the bilateral filter by changing the input parameters.

6.4. What are the pros and cons of filtering?

Next Contents

- Basic camera theory
 - Pinhole camera model
- Basic image transformation

