



# Lecture 1: Basic of Algorithm & Computational Complexity

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# 저작권 안내

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**(주)업스테이지가 제공하는 모든 교육 콘텐츠의 지식재산권은  
운영 주체인 (주)업스테이지 또는 해당 저작물의 적법한 관리자에게 귀속되어 있습니다.**

콘텐츠 일부 또는 전부를 복사, 복제, 판매, 재판매 공개, 공유 등을 할 수 없습니다.

유출될 경우 지식재산권 침해에 대한 책임을 부담할 수 있습니다.

**유출에 해당하여 금지되는 행위의 예시는 다음과 같습니다.**

- 콘텐츠를 재가공하여 온/오프라인으로 공개하는 행위
- 콘텐츠의 일부 또는 전부를 이용하여 인쇄물을 만드는 행위
- 콘텐츠의 전부 또는 일부를 녹취 또는 녹화하거나 녹취록을 작성하는 행위
- 콘텐츠의 전부 또는 일부를 스크린 캡쳐하거나 카메라로 촬영하는 행위
- 지인을 포함한 제3자에게 콘텐츠의 일부 또는 전부를 공유하는 행위
- 다른 정보와 결합하여 Upstage Education의 콘텐츠임을 알아볼 수 있는 저작물을 작성, 공개하는 행위
- 제공된 데이터의 일부 혹은 전부를 Upstage Education 프로젝트/실습 수행 이외의 목적으로 사용하는 행위

# 강사 소개

## 김수경

現 이화여자대학교 인공지능융합전공 소속 조교수

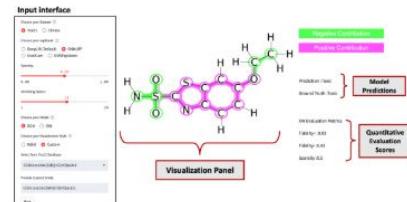


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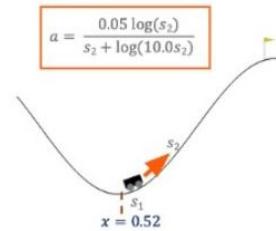


## 관심 연구 분야

- Explainable AI (XAI)
- NLP and Medical AI
- AI for Science
- RL based Optimization



Explainable AI



RL based Optimization

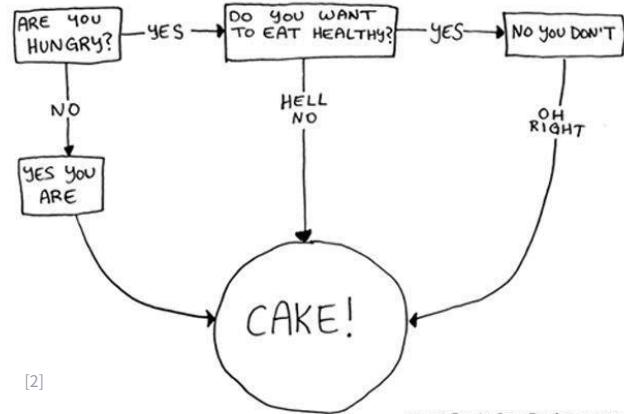
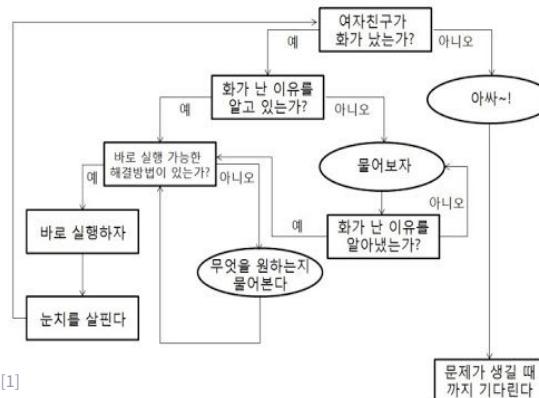
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# Algorithms & Complexity

# What is Algorithm?

- Computational procedure to solve a problem

여자친구와 싸웠을 때 알고리즘



[1] <https://blog.naver.com/sumr2002/220359932991>

[2] <http://jokesandfun.de/infographic/the-cake-is-a-factflowchart/>

# Efficiency of an Algorithm

Do you like fast/efficient computer program, or slow one?



[1]



[2]



[3]

Of course, we always expect our computers to do their jobs **most efficiently!**

[1] <https://www.vecteezy.com/photo/20621853-stressed-and-overworked-businessman>

[2] [https://stock.adobe.com/search?k=smashed+computer&asset\\_id=52541504](https://stock.adobe.com/search?k=smashed+computer&asset_id=52541504)

[3] <https://www.istockphoto.com/kr/%EC%82%AC%FC%A7%84/%EC%97%AC%EC%9F%90-%EC%BB%B4%ED%93%A8%ED%84%B0-gm118986833-12293508>

# Computation Complexity

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- Cost of algorithm = Sum of operation costs
- Model of computation specifies
  - What **operations** an algorithm is allowed to use
  - Cost (time, space) of each operation
- Execution costs
  - **Time** complexity of a program: how much **time**?
  - **Space** complexity of a program: how much **memory**?

# Measuring Time Complexity

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- **Measure execution time** in seconds using a client program (e.g., time module)
  - (+) Easy to measure
  - (+) Gives actual time
  - (-) Large amounts of time might be required.
  - (-) Results depend on lots of factors (machine, compiler, data...)
- Count the number of operations **in terms of input size  $N$** 
  - (+) Machine independent.
  - (+) Gives algorithm's **scalability**.
  - (-) Tedious to compute...
  - (-) Does not give actual time.
- Fortunately, we care only about asymptotic behavior (with a very large  $N$  – Big Data!)

# Elementary School Algorithm

## Example: Integer multiplication

- Input: two  $N$ -digit numbers  $x, y$
- Output: product of  $x$  and  $y$
- Primitive operations allowed:
  - Add 2 single-digit numbers
  - Multiply 2 single-digit numbers

$$\begin{array}{r} 233 \\ 5678 \\ \times 1234 \\ \hline 22712 \\ 17034 \\ 11356 \\ 5678 \\ \hline 7006652 \end{array}$$

# Elementary School Algorithm

## Example: Integer multiplication

- Input: two  $N$ -digit numbers  $x, y$
- Output: product of  $x$  and  $y$
- Primitive operations allowed:
  - Add 2 single-digit numbers
  - Multiply 2 single-digit numbers

How many primitive operations used?

The diagram illustrates the long multiplication of two 4-digit numbers, 233 and 5678, by 1234. The result is 7006652. The process is broken down into primitive operations:

- For each row:**  $N$  multiplications (up to  $N-1$  additions).
- $N$  rows**: In total,  $N(2N-1)$  operations (up to  $N^2$  additions).

**Total operations  $\leq 3N^2$**

233	
5678	
× 1234	
<hr/>	
22712	
17034	
11356	
5678	
<hr/>	
7006652	

# Software Engineer's Example

```
def linear_search(list, value):
    for i in range(len(list)):
        if list[i] == value:
            return i
    return -1
```

```
def selection_sort(list):
    for i in range(len(list)):
        smallest = i
        for j in range(i+1, len(list)):
            if list[j] < list[smallest]:
                smallest = j
        list[i], list[smallest] = list[smallest],
list[i]
```

Let's denote `len(list)` as  $N$ :

Operation	Count
<code>==</code>	1 to $N$

Operation	Count
<code>smallest = i</code>	$N$
<code>&lt;</code>	$(N^2 - N)/2$
<code>smallest = j</code>	0 to $(N^2 - N)/2$
<code>swap</code>	$N$

# Big O Notation

*How to Characterize Time Complexity more **formally** and **simply**?*

# Simplification of Time Complexity

## 1. We care only about the **worst-case** performance!

← because we do not know what input data we will get in advance.

Operation	Count
smallest = i	$N$
<	$(N^2 - N)/2$
smallest = j	$\cancel{O} \text{ to } (N^2 - N)/2$
swap	$N$

# Simplification of Time Complexity

2. Focus only on a single operation with the **highest order** of growth (=most expensive).

- There may be multiple good choices. Then, just choose any of them.

Operation	Count
smallest = i	$\cancel{N}$
<	$(N^2 - N)/2$
smallest = j	$(N^2 - N)/2$
swap	$\cancel{N}$

# Simplification of Time Complexity

## 3. Remove lower order terms.

← They do not really matter, since the higher order one dominates.

Operation	Count
smallest = i	<i>ignored</i>
<	<i>ignored</i>
smallest = j	$(N^2 \cancel{N})/2$
swap	<i>ignored</i>

# Simplification of Time Complexity

## 4. Remove constants.

← We have already thrown away information at step 2. At this stage, constants are not meaningful.

Operation	Count
smallest = i	<i>ignored</i>
<	<i>ignored</i>
smallest = j	$(N^2)$ 
swap	<i>ignored</i>

Worst-case order of growth:  $N^2$

# Formal Definition

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- If a function  $T(N)$  has its **order of growth** less than or equal to  $f(N)$ , we write this as  $T(N) \in O(f(N))$ , where  $O$  is called Big-O notation.
- More mathematically,  $T(N) \in O(f(N))$  means that there exists a positive constant  $c$  such that  $T(N) \leq c f(N)$  for all values of  $N$  greater than some positive  $N_0$  (*i.e.*, large  $N$ ).

# Example

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- $T(N) = N^2 + 5N^5$
- $T(N) \in O(N^5)$ ?
  - That is, is there a positive constant  $c$  such that  $N^2 + 5N^5 \leq cN^5$  for large  $N$ ?
  - Yes!
    - $N^2 + 5N^5 < N^5 + 5N^5 = 6N^5$
    - With  $c=6$ , it holds.
- $T(N) \in O(N^7)$ ?
  - That is, is there a positive constant  $c$  such that  $N^2 + 5N^5 \leq cN^7$  for large  $N$ ?
  - Yes!
    - $N^2 + 5N^5 < N^5 + 5N^5 = 6N^5 < 6N^7$
    - With  $c=6$ , it holds.

## Example(cont'd)

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- $T(N) = N^2 + 5N^5$
- $T(N) \in O(N^4)$ ?
  - That is, is there a positive constant  $c$  such that  $N^2 + 5N^5 \leq cN^4$  for large  $N$ ?
  - No 😞
    - Even if we set very large  $c$ , still for  $N > c$ ,  $5N^5 > N^5 > cN^4$ .
- We are usually interested in the **tightest** order; in this example,  $O(N^5)$ .

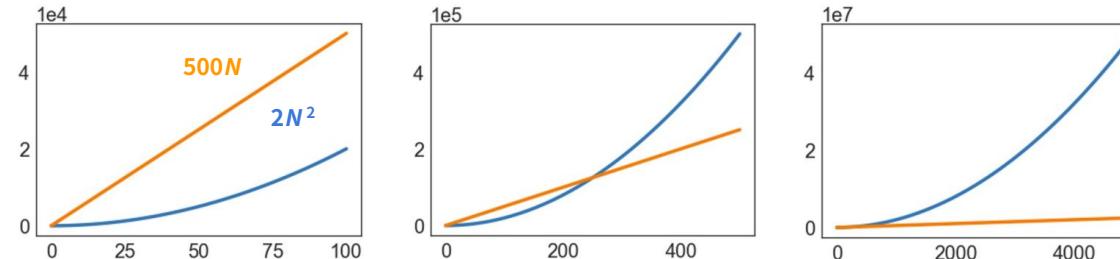
## Exercise(cont'd)

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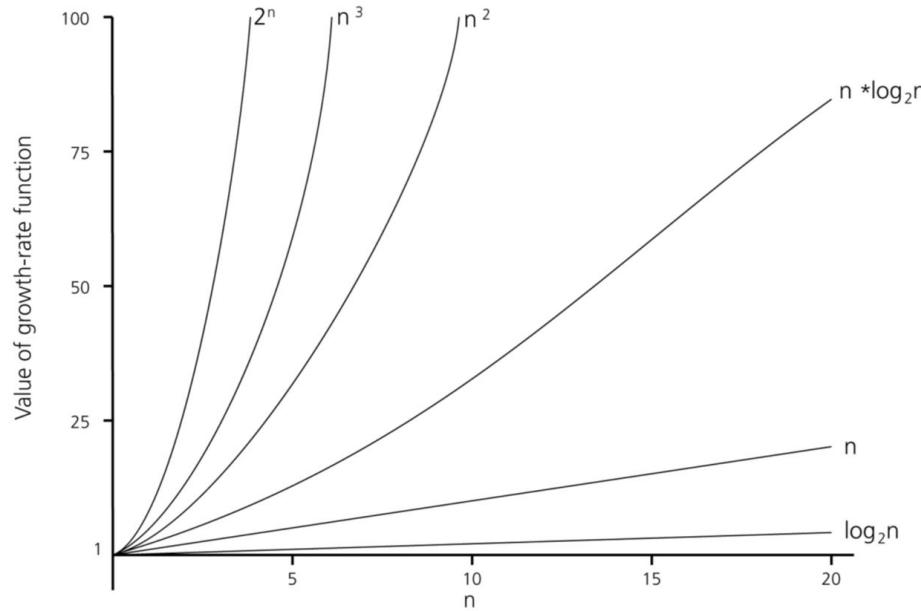
- $T(N) = 2N^2 + 3N$   $O(N^2)$
- $T(N) = 1/N + 100$   $O(1)$
- $T(N) = 100\cos(N) + 50N^2$   $O(N^2)$
- $T(N) = \log N + 2N$   $O(N)$
- $T(N) = 2^N + N^2$   $O(2^N)$

# What is Important for Asymptotic Analysis?

- Compare the two algorithms below:
  - Algo1 requires  $2N^2$  operations, while
  - Algo2 requires  $500N$  operations.
  - Algo1 is faster than Algo2 for a small  $N$ , but becomes much slower for a very large  $N$ .
  - What is important? **How fast function is growing!**
- Order of growth:



# Asymptotic Analysis





# Building intelligence for the future of work