



# **Introduction:**

## **Course Logistics and Complexity Review**

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**CSC263-Fall 2017**  
**Lecture 1**

# Today

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- About the Course
  - Why CSC263?
  - What CSC263 is about?
  - How to do well in CSC263?
- Logistics
- Review
  - Asymptotic notations (Lower bound, Upper bound, tight bound)
  - Algorithm complexity
    - Worst case, Best case, Average case

# Why CSC263?

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- ~~To graduate.~~
- To be a good programmer.
- To land a job.
- To develop excellent software in your start-up.

# Why CSC263?

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- The areas that data structures are applied extensively:

Compiler Design

Operating System

Database  
Management System

Graphics

Simulation

Artificial Intelligence

# Background (Required)

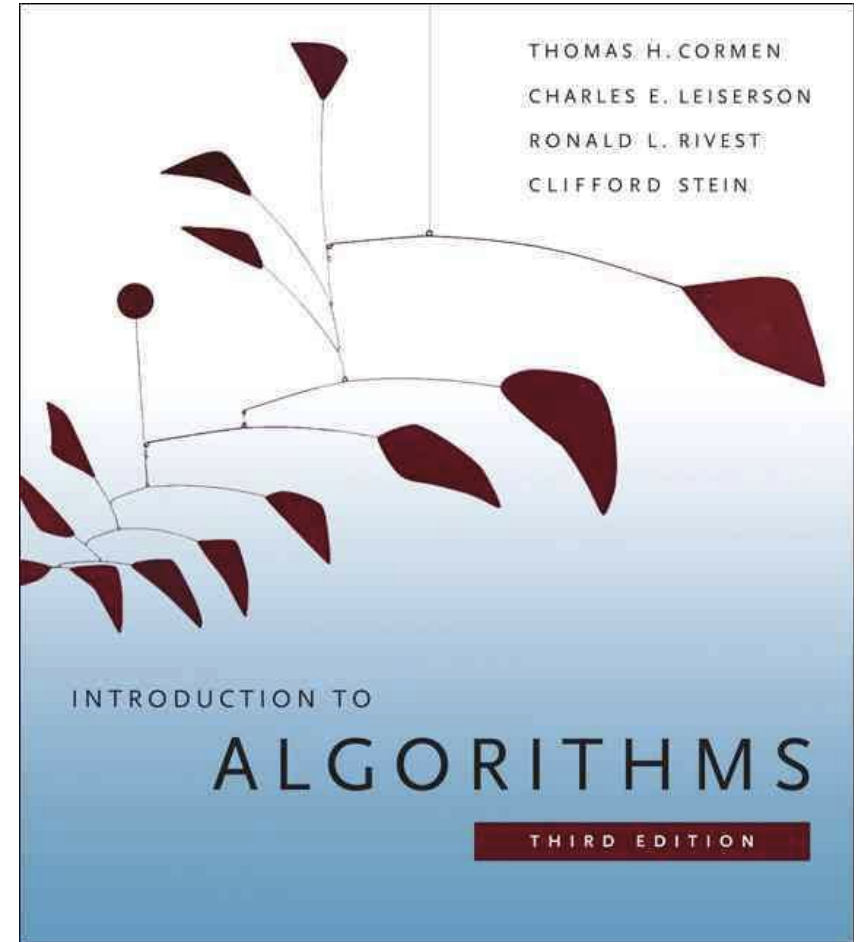
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- Theory of computation
  - Inductions
  - Recursive functions, Master Theorem
  - Asymptotic notations.
- Probability theory
  - Probabilities and counting
  - Random variables
  - Distribution
  - Expected value

# Logistics -Textbook

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- Introduction to Algorithms, Third Edition by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest and Clifford Stein.
- **Recommended book:**  
  
Data Structures and Algorithms in C++  
by Michael T. Goodrich, Roberto Tamassia, David M. Mount



# Logistics – course page

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- Course webpage:  
<http://www.cs.toronto.edu/~fpanahi/2017-fall-csc263.html>
- Course forum:  
<https://piazza.com/utoronto.ca/summer2017/csc263/home>  
All course materials and announcements will be posted in Piazza.  
We will be monitoring the course forum regularly to answer your questions.
- All assignment submissions will be done electronically, using the [Markus](#) system. Submission instruction will be provided in Piazza.

# Logistics – Sections

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- This course is offered in two sections.

Section	LEC0101, LEC2003	LEC0201, LEC2000, LEC2201
Lectures	Wed 12:00-14:00 (BA 1190)	Wed 15:00-17:00 (EM 001)
Tutorials	Fri 10:00-11:00 (BA 1190)	Fri 13:00-14:00 (EM 001)
Office Hours: By appointment	Mon 09:00 - 10:30 (BA 3219)	Mon 10:30 - 12:00 (BA 3219)

- The first tutorial will be on 15 Sep. Friday.
- For TAs' contact info, see Piazza.



# Logistics - Course Mark Composition

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- **Assignments:** 40%
  - 4 Assignments 10% each.
  - **Assignment 0:** nothing but **making a group** of two people in Markus. Deadline: Sep 20
  - **Assignment 1:** The handout will be posted on Sep 16 in Piazza.
- **Midterm Test:** 20%, Oct 25, 12:00 – 13:00 (There is a small possibility of change)  
If you cannot make it send an email to the instructor along with your document of proof before Sep 18.
- **Final Exam:** 40%, time TBA in Exam Period
- **Participation:** 2% bonus point

# Participation

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- We encourage you to participate in the lectures, tutorials and also office hours when you have questions.
  - Actively engage in lectures and tutorials.
  - Answer the questions in Piazza.
  - Win the class contests.

# Small Quizzes

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- There will be small quizzes at the beginning of some lectures to review the materials.
- You can win little prizes!



# How to do well in CSC263?

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- 1- Be interested! 😊
- 2- Solve as many exercises from the text-book as you can.
- 3- Give us feedback for any improvement.

# Quick Survey

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- Have you taken CSC263 or CSC265 before?
- Are you familiar with
  - Running time complexity of algorithms
    - Worst case, Average case, Best case
  - Asymptotic notations?
    - Big O, Big  $\Omega$ ,  $\Theta$
- Probability theory
  - Sample space
  - Expected value

# Let's get started!

**READING Assignments:** Chapters 2, 3; Sections 4.5, 5.1, 5.2.

# What CSC263 is about?

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- **Abstract Data Type (ADT):** Set of **objects** together with set of **operations** on these objects.
  - **Example: Stack**
    - Objects:** lists (or sequences)
    - Operations:** PUSH(S, v), POP(S), ISEMPTY(S)
- ADT's important for specification.
- provide modularity and reuse since **usage is independent of implementation**

# What CSC263 is about?

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- **Data Structure:** implementation of an ADT, a way to represent objects, and algorithms for every operation.

**Example: stack implementations.**

a) **Linked list** (keep pointer to head).

**ISEMPTY:** test head == None.

**PUSH:** insert at front of list.

**POP:** remove front of list (if not empty)

b) **Array with counter** (size of stack).

**ISEMPTY:** test counter == 0.

**PUSH:** insert at index counter, increment counter.

**POP:** (if not empty) decrement counter, remove from index counter.



# What CSC263 is about?

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- Analysis:
  - Correctness of algorithms
  - Run time complexity analysis

# Algorithm Analysis (Review)

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**Complexity:** amount of resources required by an algorithm, measured as a function of input size.

**Resource:** running time or memory (space), usually.

Why analyse complexity?

For comparison, e.g., choose between different implementations.

# Input size

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**Input size** is problem-dependent. Examples:

- For numbers, "size" = number of bits.
  - For lists, "size" = number of elements.
  - For graphs, "size" = number of vertices.
- 
- measure must be roughly proportional to true bit-size (no. of bits required to fully encode input).
  - In practice, allow ourselves to use  $size([a_1, a_2, \dots, a_n]) = n$ , when it is really  $size(a_1) + \dots + size(a_n)$  proportional to  $n$  only if each  $a_i$  has constant size.

# Algorithm Analysis (Review)

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For a **problem P** with **input size n** there are two algorithms.

Algorithm A

Time complexity :  $T_A(n)$

Algorithm B

Time complexity :  $T_B(n)$

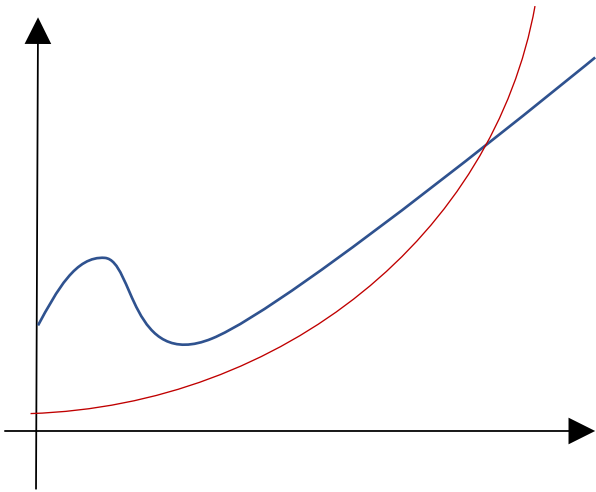
Which algorithm is better?

How do you compare two functions?

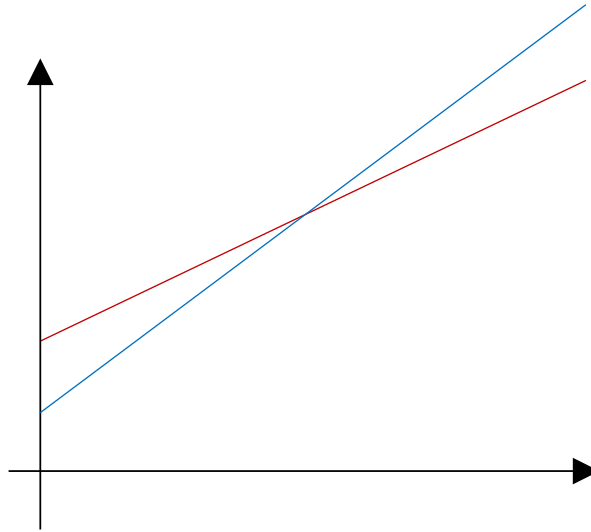
# Algorithm Analysis (Review)

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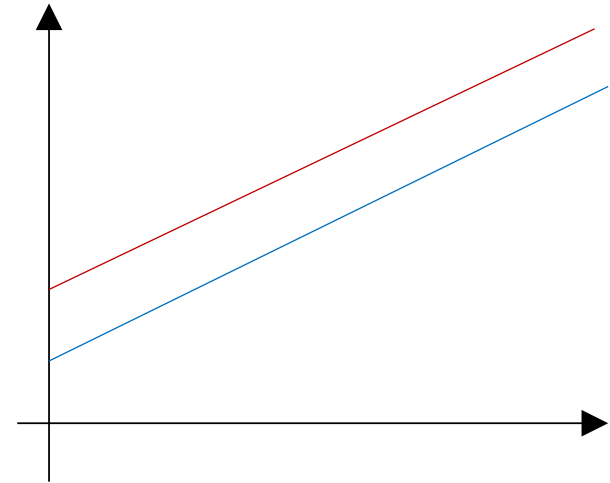
How do you compare two functions?



A



B



C

# Algorithm Analysis (Review)

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How do you compare two numbers?

$$\begin{array}{l} 2 < 3 \\ 40 < 55 \end{array}$$

Upper bound and lower bound for variables:

$$15 < x < 45$$

$$20 \leq x \leq 30$$

$$22 \leq x \leq 22$$

$x$  is 22

# Algorithm Analysis (Review)

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How would you compare two functions?

We want to define something similar for functions  $<_f$ ,  $>_f$ ,  $=_f$ ,  $\leq_f$ ,  $\geq_f$

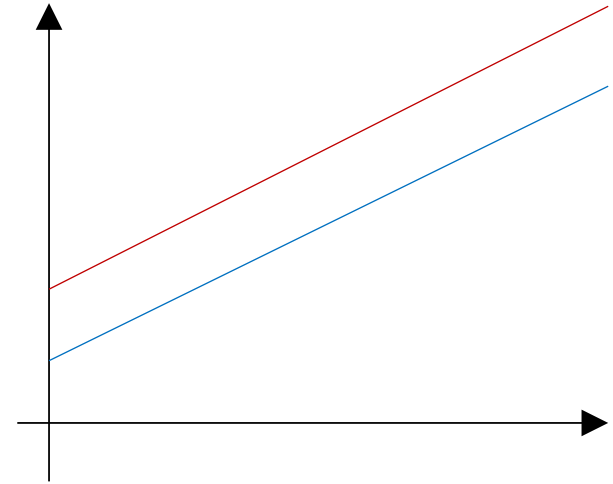
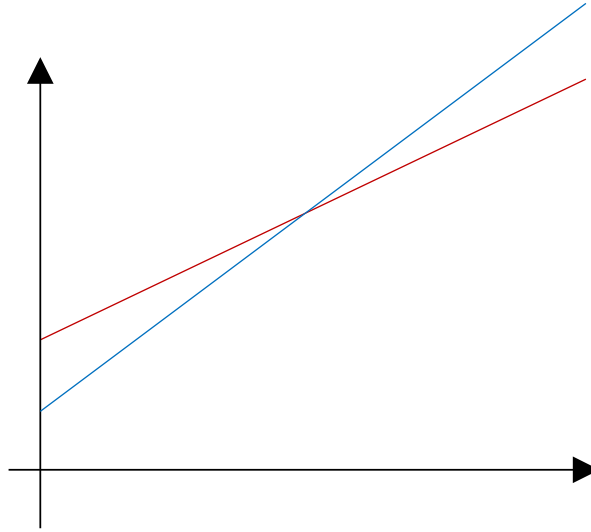
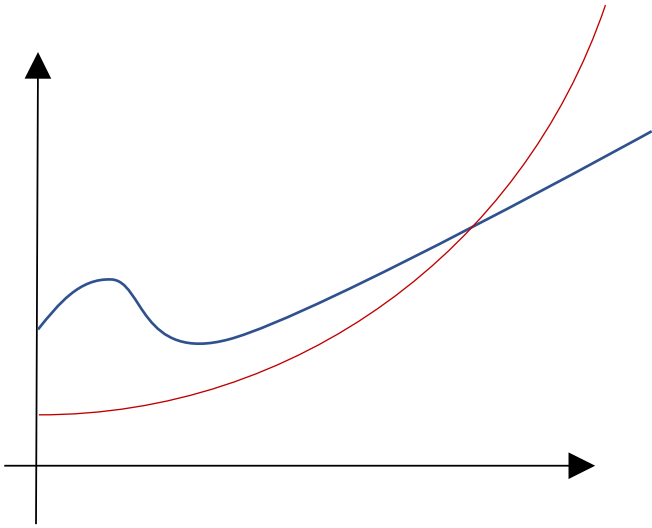
$$100x <_f 10x^2$$

Upper bound and lower bound for functions:

$$\begin{aligned}x &<_f f(x) <_f x^3 \\ \log x &\leq_f f(x) \leq_f x^2 \\ x \log x &\leq_f f(x) \leq_f x \log x \\ f(x) &= x \log x\end{aligned}$$

# Algorithm Analysis (Review)

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We **only** care about

1. Large values of  $x$
2. Growth of functions



# Asymptotic notations

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$$f(n) \leq_f g(n) \text{ or } f(n) \in O(g(n))$$

$$\rightarrow \exists n_0, c_0 > 0 \quad \text{such that for } n > n_0$$

$$0 \leq f(n) \leq c_0 g(n)$$

$$f(n) <_f g(n) \text{ or } f(n) \in o(g(n))$$

$$\rightarrow \forall c_0 > 0 \exists n_0 > 0 \quad \text{such that for } n > n_0$$

$$0 \leq f(n) < c_0 g(n)$$

$$f(n) \geq_f g(n) \text{ or } f(n) \in \Omega(g(n))$$

$$\rightarrow \exists n_0, c_0 > 0 \quad \text{such that for } n > n_0$$

$$f(n) \geq c_0 g(n) \geq 0$$

$$f(n) >_f g(n) \text{ or } f(n) \in \omega(g(n))$$

$$\rightarrow \forall c_0 > 0 \exists n_0 > 0 \quad \text{such that for } n > n_0$$

$$f(n) > c_0 g(n) \geq 0$$

$$f(n) =_f g(n) \text{ or } f(n) \in \Theta(g(n))$$

$$\rightarrow \exists n_0, c_1, c_2 > 0 \quad \text{such that for } n > n_0$$

$$0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n)$$

# Asymptotic notations

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## Examples:

$$10n^2 + 3 \in O(n^2) \in O(n^3) \in O(n^4)$$

$$10n^2 + 3 \in o(n^3) \in o(n^4)$$

$$10n^2 + 3 \in \Omega(n^2) \in \Omega(n) \in \Omega(1)$$

$$10n^2 + 3 \in \omega(n) \in \omega(1)$$

$$10n^2 + 3 \in \Theta(n^2)$$

$$40n^2 + 10n \log 5n + 100\sqrt{n} \log n + 5n + 200 \in \Theta(n^2)$$

# Asymptotic notations

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## Order of growth of some common functions

$$O(1) < O(\log^* n) < O(\log n) < O(\sqrt{n}) < O(n) < O(n \log n) < O(n^2) < O(n^3) < O(2^n)$$

$$\log^* n = \min \{ i \geq 0 : \log^{(i)} n \leq 1 \}$$

The iterated logarithm is a very slowly growing function:

$$\log^* 2 = 1$$

$$\log^* 4 = 2$$

$$\log^* 16 = 3$$

$$\log^* 65536 = 4$$

$$\log^* (2^{65536}) = 5$$

# Asymptotic notations (Upper bound and Lower bound)

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Let  $T(n) = \pi n \lg n + 2^{12} n - \sqrt{n} \lg n$

What is the best lower bound and upper bound to  $T(n)$  that you find in the options listed here.

$$\Omega(1)$$

$$O(1)$$

$$\Omega(\lg n)$$

$$O(\lg n)$$

$$\Omega(\lg^2 n)$$

$$O(\lg^2 n)$$

$$\Omega(\sqrt{n} \lg n)$$

$$O(\sqrt{n} \lg n)$$

$$\Omega(n)$$

$$O(n)$$

$$\Omega(n\sqrt{n})$$

$$O(n\sqrt{n})$$

$$\Omega(n^2)$$

$$O(n^2)$$

# Coding examples

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## Example 1:

```
for (  $i = 0 ; i < n ; i++$  )  
     $m += i$ ;
```

## Example 2:

```
for (  $i = 0 ; i < n ; i++$  )  
    for(  $j = 0 ; j < n ; j++$  )  
         $sum[i] += entry[i][j]$ ;
```

# Coding examples

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## Example 3:

```
for (  $i = 0 ; i < n ; i++$  )  
    for(  $j = 0 ; j < i ; j++$  )  
         $m += j$ ;
```

## Example 4:

```
 $i = 1$ ;  
while (  $i < n$  ) {  
     $tot += i$ ;  
     $i = i * 2$ ;  
}
```

# Coding examples

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## Example 5:

```
i = n;  
while (i > 0) {  
    tot += i;  
    i = i / 2;  
}
```

## Example 6:

```
for ( i = 0 ; i < n ; i + + )  
    for( j = 0 ; j < n ; j + + )  
        for( k = 0 ; k < n ; k + + )  
            sum[i][j] += entry[i][j][k];
```

# Coding examples

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## Example 7:

```
for (  $i = 0 ; i < n ; i++$  )  
    for(  $j=0 ; j < \sqrt{n} ; j++$  )  
         $m += j$ ;
```

## Example 8:

```
for (  $i = 0 ; i < n ; i++$  )  
    for(  $j = 0 ; j < \sqrt{995} ; j++$  )  
         $m += j$ ;
```



# Coding examples

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## Example 9:

```
for ( i = 0 ; i < n ; i + + )  
{  
    m += j;  
    m += j;  
    m += j;  
    ...  
    m += j; // 31 times  
}
```

# Complexity Review - Example

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**Example:** Input  $A$ ,  $A.length = n$

```
1:    $i = NIL$ 
2:   for  $j = 0$  to  $n - 1$  do
3:       if  $A[j] = v$  then
4:            $i = j$ 
5:       return  $i$ 
7:   return  $i$ 
```

# Running time, Space complexity

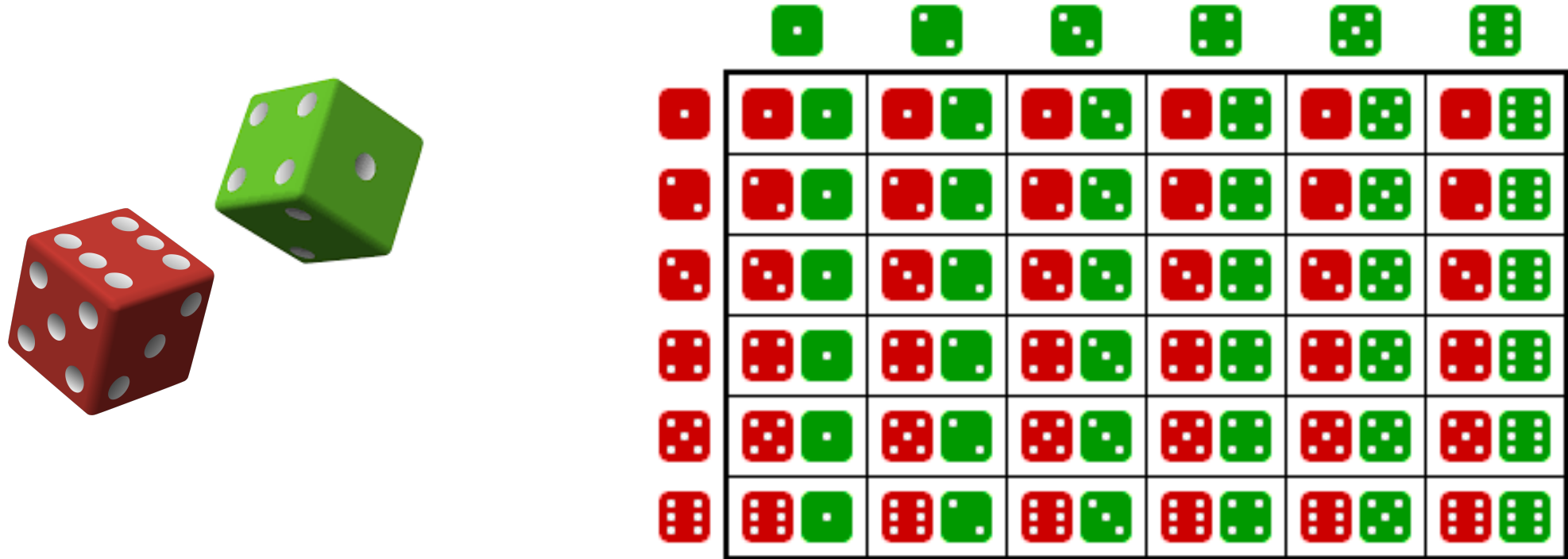
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- **Best Case:** Minimum complexity over the sample space
- **Average Case:** Expected value over the sample space by considering the probability distribution over inputs
- **Worst case:** Maximum complexity over the sample space
- What was **sample space**?!

# Probability Review

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- **Outcome:** a possible result of an **experiment (Algorithm)**. (Rolling 2 dice and getting 1,1)
- **Event:** a set of one or more outcomes. (Rolling 2 dice and getting only even numbers)
- **Sample Space**  $S$  is a set of elementary events. (Sample space for rolling 2 dice)



# Complexity Review – Sample Space

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- Any problem has some **input**.
- **Sample space** for a problem is the set all possible inputs that effects on the behavior of the algorithm.

# Complexity Review

**Example:** Input  $A$ ,  $A.length = n$

```
1:    $i = NIL$ 
2:   for  $j = 0$  to  $n - 1$  do
3:       if  $A[j] = v$  then
4:            $i = j$ 
5:       return  $i$ 
7:   return  $i$ 
```

- **Sample space:** infinitely many inputs!
- behavior of the algorithm determined by only one factor : **position of  $v$  inside  $A$ .**

*$v$  occurs at position 0 in  $A$ ,*

*$v$  occurs at position 1 in  $A$*

*...*

*$v$  occurs at position  $n - 1$  in  $A$ ,*

*$v$  does not occur in  $A$ .*

So pick "representative" inputs (one for each execution time),

**$S_n = \{ (A, v) : A = [0, 1, 2, \dots, n - 1] \text{ and } v = 0, 1, 2, \dots, n \}.$**

# Running time

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- For algorithm A,  $t(x)$  represents number of "steps" executed by A on input  $x$  with size  $n$ .
- **Best-case** running time  $T(n) = \min \{t(x) : x \text{ is an input of size } n\}$
- **Worst-case** running time  $T(n) = \max \{t(x) : x \text{ is an input of size } n\}$   
(Mostly useless!)

**Example:** Input A

```
1:    $i = NIL$ 
2:   for  $j = 0$  to  $n - 1$  do
3:       if  $A[j] = v$  then
4:            $i = j$ 
5:       return  $i$ 
7:   return  $i$ 
```

**Best-Case:**

$O(1), \Omega(1), \Theta(1)$

**Worst Case:**

$O(n), \Omega(n), \Theta(n)$

# Average-Case running time

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- **Average Case:** Expected value over the sample space by considering the probability distribution over inputs.

$t_n$  be a random variable that denotes the number of comparisons executed (Line #3)

$$E(t_n) = \sum_{(A,v) \in S_n} t_n(A,v) \times P[(A,v)]$$

$\Pr[(A,n)] = p,$	$v$ not in $A$ as special
$\Pr[(A,v)] = (1-p)/n$	other cases equally likely



# Average-Case running time

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Let  $t_n(A, v)$  be the number of execution of line 3 for input  $A, v$ .

$$t_n(A, v) = \begin{cases} v + 1 & \text{if } 0 \leq v \leq n - 1 \\ n & \text{if } v = n \end{cases}$$

$$E(t_n) = \sum_{(A,v) \in S_n}^n t_n(A, v) \times P[(A, v)] = n \times p + \sum_{v=0}^{n-1} (v + 1) \times (1 - p)/n$$

$$= np + \frac{1-p}{n} \sum_{v=1}^n v = np + (1-p)(n+1)/2 = (n+1+np-p)/2$$

Average case:  $O(n), \Omega(n), \Theta(n)$

Questions?