



COMP 1002

Logic for Computer Scientists

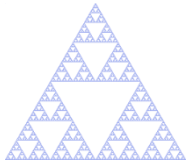
Lecture 30



Tower of Hanoi game

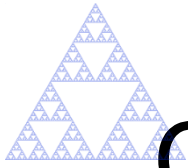


- Rules of the game:
 - Start with all disks on the first peg.
 - At any step, can move a disk to another peg, as long as it is not placed on top of a smaller disk.
 - Goal: move the whole tower onto the second peg.
- Question: how many steps are needed to move the tower of 8 disks? How about n disks?



Recurrence relations

- **Recurrence:** an equation that defines an n^{th} element in a sequence in terms of one or more of previous terms.
 - Think of $F(n) = s_n$ for some sequence $\{s_n\}$
 - $H(n) = 2H(n - 1) + 1$
 - $F(n) = F(n - 1) + F(n - 2)$
- A **closed form** of a recurrence relation is an expression that defines an n^{th} element in a sequence in terms of n directly.
 - Often use recurrence relations and their closed forms to describe performance of (especially recursive) algorithms.



Closed forms of some sequences

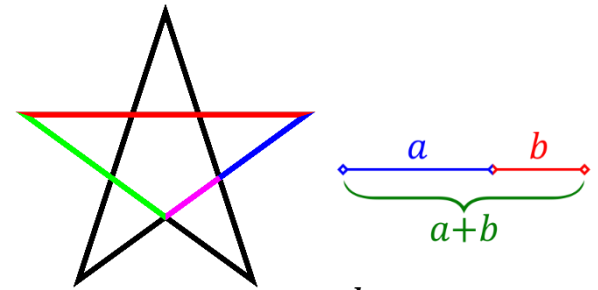
- Arithmetic progression:
 - Sequence: $c, c + d, c + 2d, c + 3d, \dots, c + nd, \dots$
 - **Closed form:** $s_n = c + nd$
 - Closed forms are very useful for analysis of recursive programs, etc.
- Geometric progression:
 - Sequence: $c, cr, cr^2, cr^3, \dots, cr^n, \dots$
 - **Closed form:** $s_n = c \cdot r^n$

- Fibonacci sequence: $F(n) = F(n-1) + F(n-2)$

– Sequence: 1, 1, 2, 3, 5, 8, 13, ...

– **Closed form:** $F_n = \frac{\varphi^n - (1-\varphi)^n}{\sqrt{5}}$

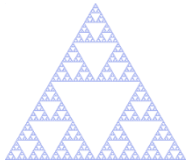
- Where φ (“phi”) is the “golden ratio”: a ratio such that $\frac{a+b}{a} = \frac{a}{b}$
- $\varphi = \frac{1+\sqrt{5}}{2}$



Tower of Hanoi game



- Rules of the game:
 - Start with all disks on the first peg.
 - At any step, can move a disk to another peg, as long as it is not placed on top of a smaller disk.
 - Goal: move the whole tower onto the second peg.
- Question: how many steps are needed to move the tower of 8 disks? How about n disks?
- Let us call the number of moves needed to transfer n disks $H(n)$.
 - Names of pegs do not matter: from any peg i to any peg $j \neq i$ would take the same number of steps.
- Basis: only one disk can be transferred in one step.
 - So $H(1) = 1$
- Recursive step:
 - suppose we have $n-1$ disks. To transfer them all to peg 2, need $H(n - 1)$ number of steps.
 - To transfer the remaining disk to peg 3, 1 step.
 - To transfer $n-1$ disks from peg 2 to peg 3 need $H(n-1)$ steps again.
 - So $H(n) = 2H(n-1)+1$ (recurrence).
- Closed form: $H(n) = 2^n - 1$.



Closed form for Tower of Hanoi

- Solving a recurrence: finding a closed form.

– Solving the recurrence $H(n)=2H(n-1)+1$

- $H(n) = 2 \cdot H(n - 1) + 1$

$$= 2(2H(n - 2) + 1) + 1 = 2^2H(n - 2) + 2 + 1$$

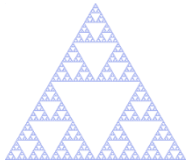
$$= 2^3H(n - 3) + 2^2 + 2 + 1$$

$$= 2^4 H(n - 4) + 2^3 + 2^2 + 2 + 1 \dots$$

– Closed form: $H(n) = \sum_{i=0}^{n-1} 2^i = 2^n - 1$

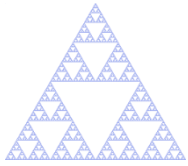
- Proof by induction

- Or by noticing that a binary number 111...1 plus 1 gives a binary number 10000...0



Solving recurrences

- So adding one more disk doubles the number of steps.
 - We say that the function defined by $H(n)$ **grows exponentially**
 - $H(n) \in O(2^n)$ (and nothing slower-growing).
 - To say “nothing slower-growing”, use symbol Ω (uppercase omega):
 $H(n) \in \Omega(2^n)$
 - To say “grows exactly like 2^n ”, use symbol Θ (uppercase theta):
 $H(n) \in \Theta(2^n)$
- Solving recurrences in general might be tricky.
 - When the recurrence is of the form $T(n) = aT(n/b) + f(n)$, there is a general method to estimate the growth rate of a function defined by the recurrence
 - Called the Master Theorem for recurrences.



Master theorem for solving recurrences

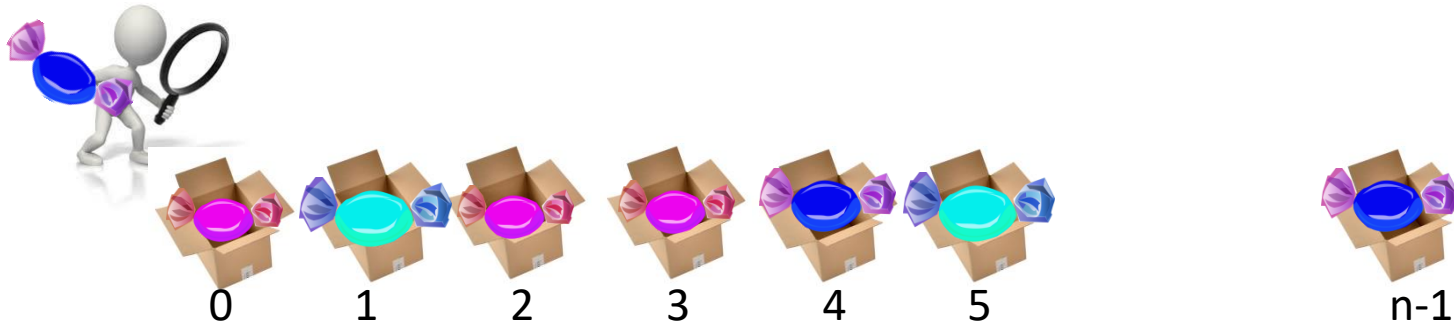
- Let $a, b, c, d \in \mathbb{R}$ such that $a \geq 1$, $b \geq 2$, $c > 0$, $d \geq 0$, and let $f(n) \in \Theta(n^c)$
- Let $T(n)$ be the following recurrence relation:
 - Base: $T(1) = d$
 - Recurrence: $T(n) = a T\left(\left\lfloor \frac{n}{b} \right\rfloor\right) + f(n)$
- Then the growth rate of $T(n)$ is:
 - If $\log_b a < c$ then $T(n) \in \Theta(f(n))$
 - If $\log_b a = c$ then $T(n) \in \Theta(f(n) \log n)$
 - If $\log_b a > c$ then $T(n) \in \Theta(n^{\log_b a})$



Analysis of algorithms

- Putting it all together:
 - Using **logic** to describe what an algorithm is doing
 - and **induction** to show that it does that correctly
 - Using **recurrence** relations to see how long it takes in the worst case.
 - With **O-notation** to talk about the time.
 - and **probabilities/expectation** to try to see how long it might take on average.

Example: search in an array



- Given:

- an array A containing n elements,
- and a specific item x

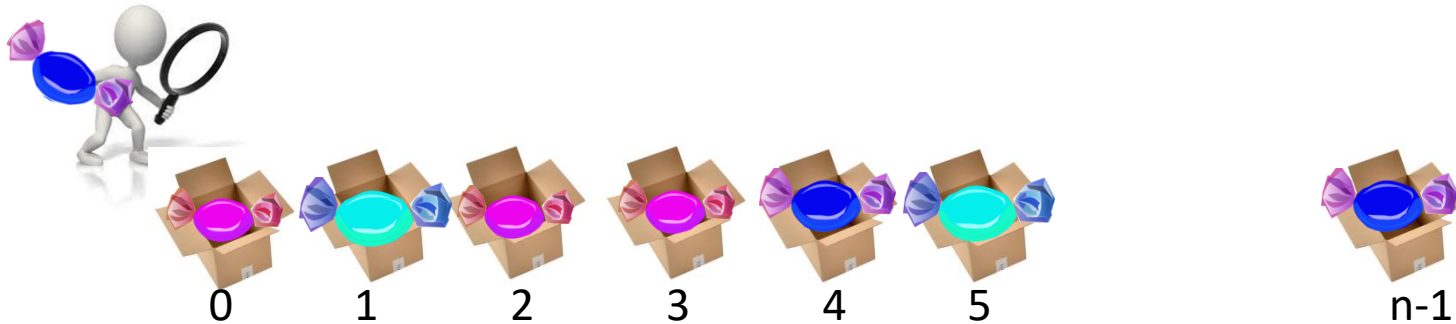


- Goal: find the index of x in A , if x is in A .

- Which box contains ? Box 4.



Example: search in an array



- Given:

- an array A containing n elements,
- and a specific item x

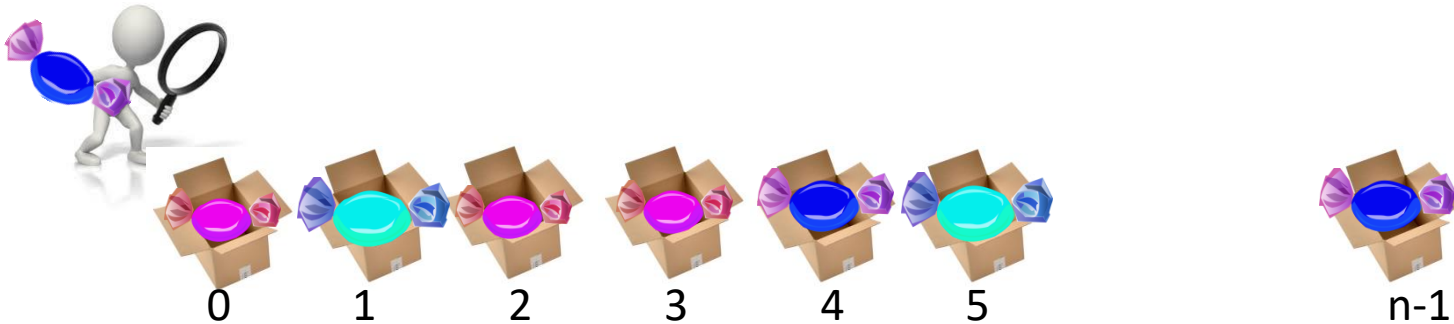


- Goal: find the index of x in A , if x is in A .

- Which box contains ? Box 4.



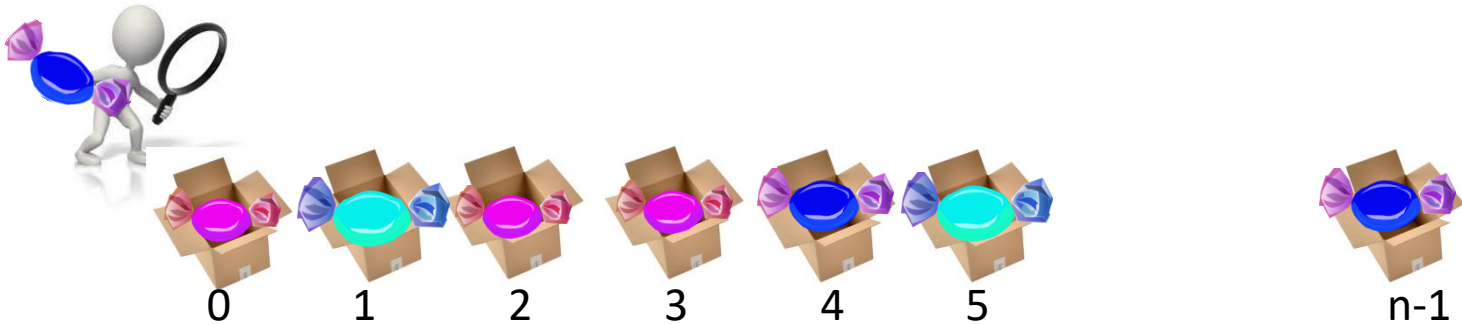
Example: search in an array



- *Precondition*: what should be true before a piece of code (or the whole algorithm) starts
 - E.g.: A is an array of numbers and A is not empty and x is a number.
- *Postcondition*: what should be true after a program (piece of code) finished.
 - E.g. If the program returned value k , then $A[k]=x$
 - or $k=-1$, if x is not in A .



Example: search in an array

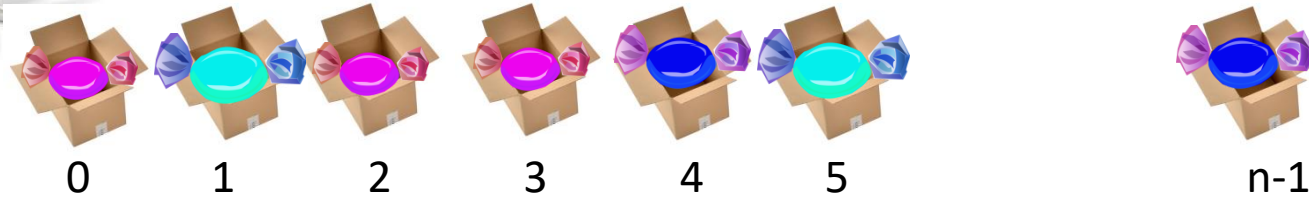


- *Precondition:* A is an array containing x
- *Postcondition:* Returned k such that $A[k]=x$





Example: search in an array



- *Precondition*: A is an array containing x

Algorithm *arraySearch*(A, x)

Input array A of n integers, number x

Output k such that $A[k]=x$

$i = 0$

$out = -1$

while $out < 0$ **do**

if $A[i] = x$ **then**

$out = i$

$i = i + 1$

return out

- *Postcondition*: Returned k such that $A[k]=x$

arraySearch algorithm

Algorithm *arraySearch*(A, x)

Input array A of n integers, number x

Output k such that $A[k]=x$

$\exists i \in \{0 \dots n - 1\} A[i] = x$

$i = 0$

$out = -1$

$\exists i \in \{0 \dots n - 1\} A[i] = x \wedge i = 0 \wedge out = -1$

while $out < 0$ **do**

if $A[i] = x$ **then**

$out = i$

$i = i + 1$

$A[out] = x$

return out

Program returned k such that $A[k]=x$

- $A = [5, 10, 8, 7]$
- $x = 8$
- $out = 2$

Loop invariant

- **Loop invariant:** a condition that is true on each iteration of the loop
 - Implied by loop precondition
 - Implies the loop postcondition
 - Implies next loop iteration is correct
- $I(k): i = k \wedge ((out = i \wedge A[out] = x) \vee (\exists j > i \ A[j] = x))$
- Guard condition: condition in the while loop
 - $G = \text{“out} < 0\text{”}$
- Loop is correct when:
 - precondition $\rightarrow I(0)$
 - for all k , $G \wedge I(k) \rightarrow I(k + 1)$
 - If k_0 is the smallest number such that $\neg G$, then $\neg G \wedge I(k_0) \rightarrow \text{postcondition}$
- **Termination:** proof that $\exists k_0$ such that after k_0 iterations G becomes false

```

$$\exists i \in \{0 \dots n - 1\} \ A[i] = x \wedge$$

$$\wedge i = 0 \wedge out = -1$$

```

```
while  $out < 0$  do  
    if  $A[i] = x$  then  
         $out = i$   
         $i = i + 1$ 
```

```
 $A[out] = x$ 
```


Proving the loop invariant

- By induction on i :
- Base case: $I(0)$

$$\begin{aligned} - \exists i \in \{0 \dots n - 1\} \ A[i] = x \wedge i = 0 \wedge \\ \wedge out = -1 \end{aligned}$$

Implies $I(0)$

$$- i = 0 \wedge ((out = 0 \wedge A[out] = x) \vee (\exists j > i \ A[j] = x))$$

- Assume $I(k)$: $i = k \wedge ((out = i \wedge A[out] = x) \vee (\exists j > i \ A[j] = x))$
- Show: if G , then $I(k+1)$: $i = k + 1 \wedge ((out = i \wedge A[out] = x) \vee (\exists j > i \ A[j] = x))$
 - $i=k+1$ because of “ $i=i+1$ ” statement
 - If $A[i]=x$, then $(out = i \wedge A[out] = x)$ holds
 - Otherwise, $(\exists j > i \ A[j] = x)$ holds.
- Otherwise, if $\neg G$, postcondition holds:
 - in this case, $(out = i \wedge A[out] = x)$ should have been true in $I(k)$, for $i=k$.
 - So $A[out]=x$

$$\begin{aligned} \exists i \in \{0 \dots n - 1\} \ A[i] = x \wedge \\ \wedge i = 0 \wedge out = -1 \end{aligned}$$

```
while out < 0 do
  if A[i] = x then
    out = i
  i = i+1
```

$$A[out] = x$$