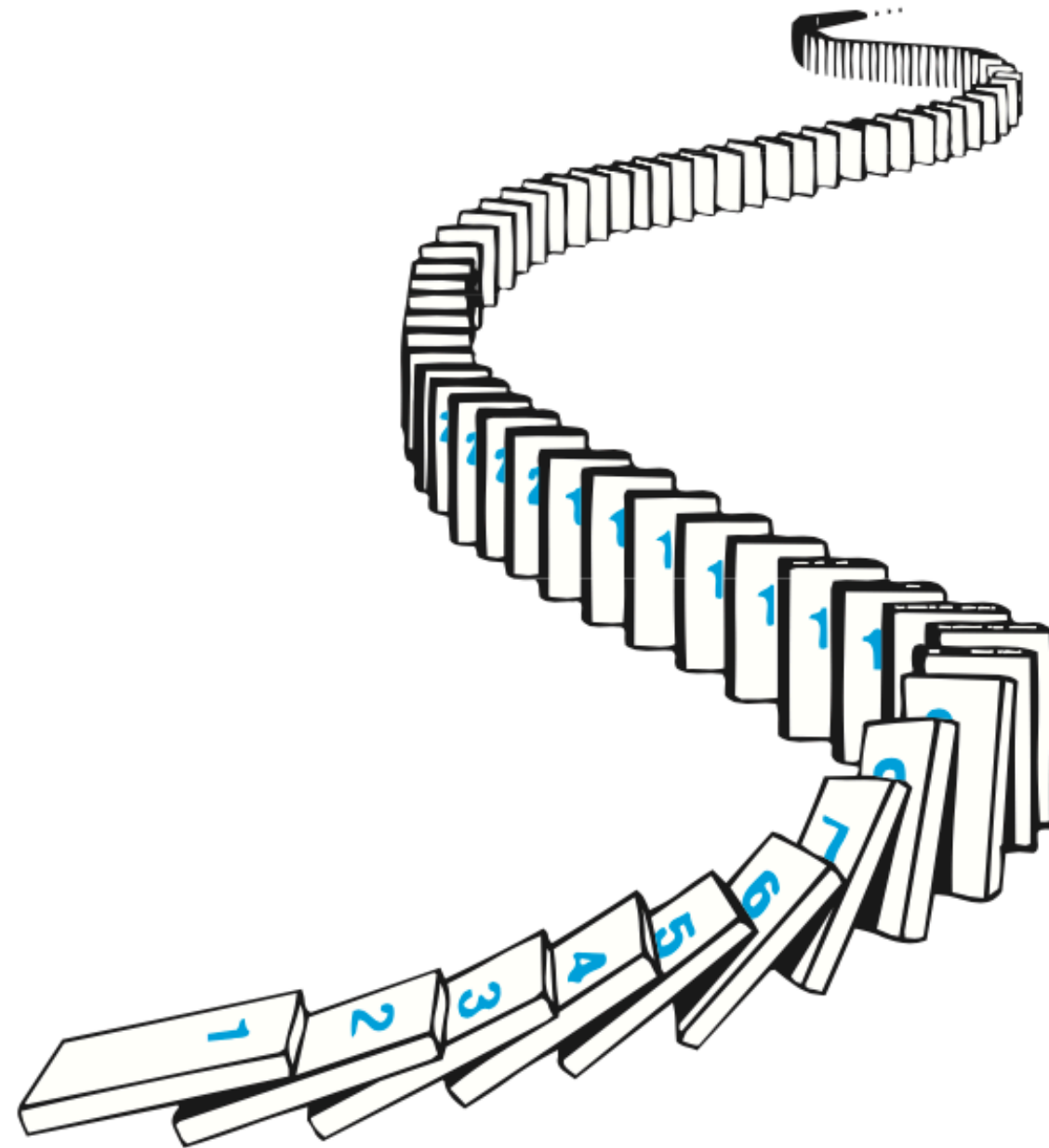
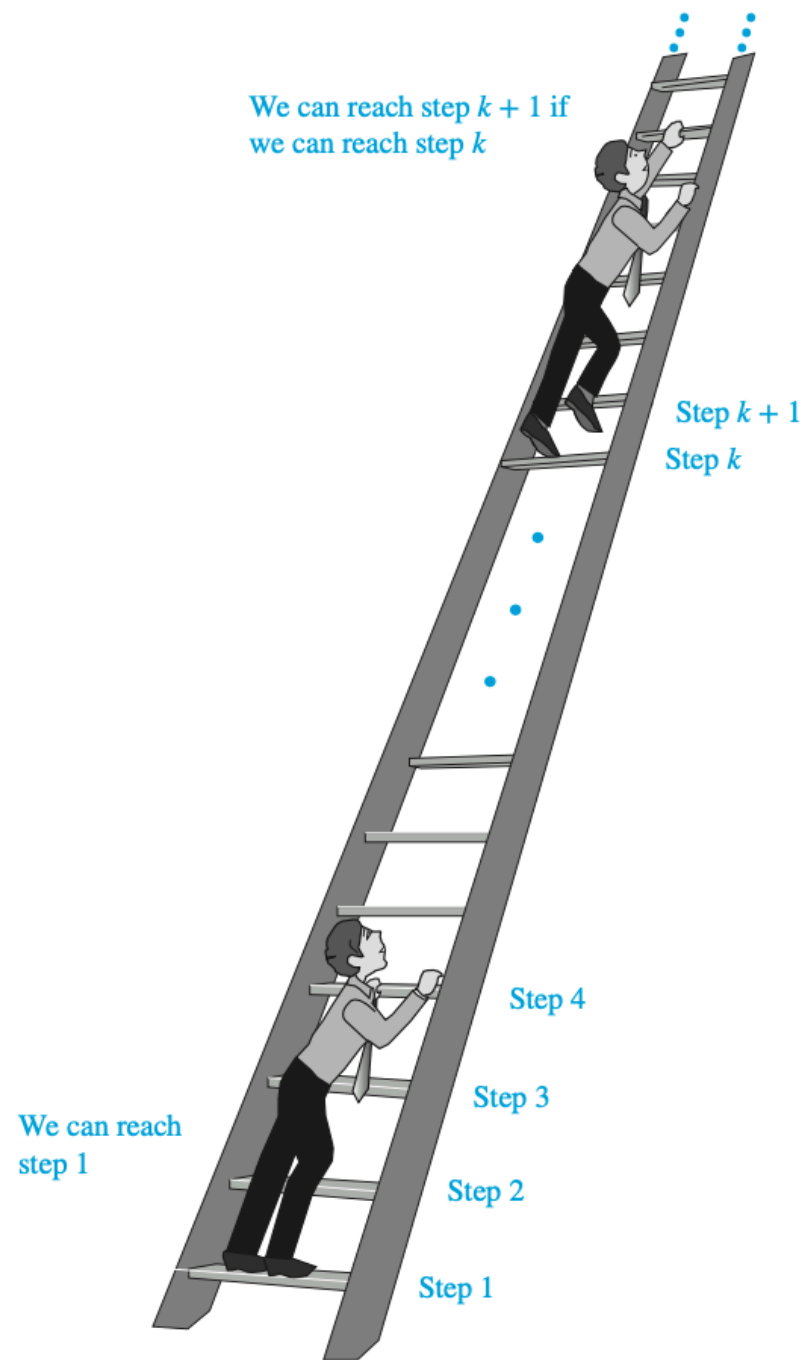


06 Induction and Recursion

CS201 Discrete Mathematics

Instructor: Shan Chen

Mathematical Induction



Mathematical Induction

Principle of Mathematical Induction

- Let $P(n)$ be a **propositional function**, i.e., $P(n)$ is either true or false when n is a specific number.
- **Principle of mathematical induction:** To prove that $P(n)$ is true for all positive integers n , we complete two steps:
 - **Basis step:** prove $P(1)$ is true
 - **Inductive step:** prove $\forall k \in \mathbb{Z}^+, P(k) \rightarrow P(k + 1)$ is true
* the assumption “ $P(k)$ is true” is called the **inductive hypothesis**
- **Q:** Why is this principle valid?
- Proof by contradiction: Assume $P(n)$ is false for some integer $n \geq 1$, then the **set S** of all positive integer n such that $P(n)$ is false is **non-empty**. Let m be the **smallest** integer in S . * *why m exists?*
We have $m \geq 2$ as $P(1)$ is true. However, since $P(m - 1)$ is true and $P(m - 1) \rightarrow P(m)$ is true, $P(m)$ must be true, contradiction!

Principle of Mathematical Induction

- **Principle of mathematical induction:** To prove that $P(n)$ is true for all positive integers n , we complete two steps:
 - **Basis step:** prove $P(1)$ is true
 - **Inductive step:** prove $\forall k \in \mathbb{Z}^+, P(k) \rightarrow P(k + 1)$ is true
- **Proof by contradiction:** Assume $P(n)$ is false for some integer $n \geq 1$, then the set S of all positive integer n such that $P(n)$ is false is non-empty. Let m be the smallest integer in S . * *why m exists?*
We have $m \geq 2$ as $P(1)$ is true. However, since $P(m - 1)$ is true and $P(m - 1) \rightarrow P(m)$ is true, $P(m)$ must be true, contradiction!
- **Well-ordering principle (axiom):** every nonempty subset of the set of positive integers has a least/minimum element.
 - this principle is equivalent to principle of mathematical induction
* *proof left as an exercise*

Example

- Show that $1 + 2 + \dots + n = n(n + 1)/2$ for any positive integer n .
- Proof by induction:
 - Let $P(n)$ be the proposition that the sum of the first n positive integers is equal to $n(n + 1)/2$.
 - **Basis step:** $P(1)$ is true, because $1 = 1(1 + 1)/2$.
 - **Inductive step:** From the inductive hypothesis, i.e., $P(k)$ is true for an arbitrary positive integer k , we need to show that $P(k + 1)$ is true, i.e., $1 + 2 + \dots + k + 1 = (k + 1)((k + 1) + 1)/2$.
$$\begin{aligned}1 + 2 + \dots + k + (k + 1) &= k(k + 1)/2 + k + 1 \\&= (k(k + 1) + 2(k + 1))/2 = (k + 1)(k + 2)/2 = (k + 1)((k + 1) + 1)/2\end{aligned}$$
 - By mathematical induction, we know that $P(n)$ is true for all positive integers n . That is, we have proven that $1 + 2 + \dots + n = n(n + 1)/2$ holds for all positive integers n .

Exercise (5 mins)

- For any positive integer n , $1 + 3 + 5 + \dots + (2n - 1) = ?$ Prove it.

- Show that $1 + 2 + \dots + n = n(n + 1)/2$ for any positive integer n .
- Proof by induction:
 - Let $P(n)$ be the proposition that the sum of the first n positive integers is equal to $n(n + 1)/2$.
 - **Basis step:** $P(1)$ is true, because $1 = 1(1 + 1)/2$.
 - **Inductive step:** From the inductive hypothesis, i.e., $P(k)$ is true for an arbitrary positive integer k , we need to show that $P(k + 1)$ is true, i.e., $1 + 2 + \dots + k + 1 = (k + 1)((k + 1) + 1)/2$.
$$\begin{aligned} 1 + 2 + \dots + k + (k + 1) &= k(k + 1)/2 + k + 1 \\ &= (k(k + 1) + 2(k + 1))/2 = (k + 1)(k + 2)/2 = (k + 1)((k + 1) + 1)/2 \end{aligned}$$
 - By mathematical induction, we know that $P(n)$ is true for all positive integers n . That is, we have proven that $1 + 2 + \dots + n = n(n + 1)/2$ holds for all positive integers n .

Exercise (5 mins)

- For any positive integer n , $1 + 3 + 5 + \dots + (2n - 1) = ?$ Prove it.
- $1 + 3 + 5 + \dots + (2n - 1) = n^2$
 - $1 + 3 = 4$, $1 + 3 + 5 = 9$, ...
- Proof by induction: (Let $P(n)$ be the above equation.)
 - **Basis step:** $P(1)$ is true, because $1 = 1^2$.
 - **Inductive step:** From the inductive hypothesis, i.e., $P(k)$ is true for an arbitrary positive integer k , we need to show that $P(k + 1)$ is true, i.e., $1 + 3 + \dots + 2(k + 1) - 1 = (k + 1)^2$.
$$\begin{aligned}1 + 3 + \dots + 2k - 1 + 2(k + 1) - 1 &= k^2 + 2(k + 1) - 1 \\&= k^2 + 2k + 2 - 1 = k^2 + 2k + 1 = (k + 1)^2\end{aligned}$$
 - By mathematical induction, we know that $P(n)$ is true for all positive integers n . That is, we have proven that $1 + 3 + 5 + \dots + (2n - 1) = n^2$ for all positive integers n .

Exercise (5 mins)

○ Prove that for any integer $n \geq 2$, $2^{n+1} \geq n^2 + 3$

○ Show that $1 + 2 + \dots + n = n(n + 1)/2$ for any positive integer n .

○ Proof by induction:

- Let $P(n)$ be the proposition that the sum of the first n positive integers is equal to $n(n + 1)/2$.
- **Basis step:** $P(1)$ is true, because $1 = 1(1 + 1)/2$.
- **Inductive step:** From the inductive hypothesis, i.e., $P(k)$ is true for an arbitrary positive integer k , we need to show that $P(k + 1)$ is true, i.e., $1 + 2 + \dots + k + 1 = (k + 1)((k + 1) + 1)/2$.

$$\begin{aligned} 1 + 2 + \dots + k + (k + 1) &= k(k + 1)/2 + k + 1 \\ &= (k(k + 1) + 2(k + 1))/2 = (k + 1)(k + 2)/2 = (k + 1)((k + 1) + 1)/2 \end{aligned}$$

- By mathematical induction, we know that $P(n)$ is true for all positive integers n . That is, we have proven that $1 + 2 + \dots + n = n(n + 1)/2$ holds for all positive integers n .

Exercise (5 mins)

- Prove that for any integer $n \geq 2$, $2^{n+1} \geq n^2 + 3$
- Proof by induction:
 - Let $P(n)$ be $2^{n+1} \geq n^2 + 3$.
 - **Basis step:** $P(2)$ is true, because $2^{2+1} = 8 \geq 7 = 2^2 + 3$.
 - **Inductive step:** From the inductive hypothesis, i.e., $P(k)$ is true for an arbitrary integer $k \geq 2$, we need to show that $P(k + 1)$ is true:
$$\begin{aligned} 2^{(k+1)+1} &= 2 \cdot 2^{k+1} \geq 2(k^2 + 3) = 2k^2 + 6 = (k + 1)^2 - 2k - 1 + k^2 + 6 \\ &= (k + 1)^2 + (k - 1)^2 + 4 \geq (k + 1)^2 + 3 \end{aligned}$$
 - By mathematical induction, $P(n)$ is true for all integers $n \geq 2$.

Another Form of Induction

- We may have another form of **mathematical induction** as follows:
 - First suppose that we have a proof that $P(1)$ is true.
 - Next suppose that we have a proof that
$$\forall k \geq 1, P(1) \wedge P(2) \wedge \dots \wedge P(k) \rightarrow P(k + 1)$$
 - Then,
$$P(1) \rightarrow P(2)$$
$$P(1) \wedge P(2) \rightarrow P(3)$$
$$P(1) \wedge P(2) \wedge P(3) \rightarrow P(4)$$
$$\dots$$
 - Iterating gives us a proof of $P(n)$ for all n

Strong Induction

- **Second principle of mathematical induction:** To prove that $P(n)$ is true for all positive integers n , we complete two steps:
 - **Basis step:** prove $P(1)$ is true
 - **Inductive step:** prove $\forall k \in \mathbf{Z}^+, P(1) \wedge \dots \wedge P(k) \rightarrow P(k+1)$ is true
** the assumption “ $P(1) \wedge P(2) \wedge \dots \wedge P(k)$ is true” is called the **inductive hypothesis***
- This is called **strong induction** or **complete induction**, while the previous principle is called **weak** or **incomplete** induction.
- In practice, strong induction is often easier to apply than its weak form, because the **inductive hypothesis is stronger**.
- However, these two forms of induction are actually **equivalent**.
 - *proof left as an exercise*

Example

- Prove that every positive integer is a power of a prime or the product of powers of primes.
- Proof:
 - **Basis step:** 1 is a power of a prime number, $1 = 2^0$.
 - **Inductive step:**

Inductive hypothesis: every positive integer that is $\leq k$ is a power of a prime or a product of powers of primes.

If $k + 1$ is a prime power, $P(k + 1)$ is true. Otherwise, $k + 1$ must be a composite, i.e., a product of two smaller positive integers, each of which is, by the **inductive hypothesis**, a power of a prime or the product of powers of primes. Therefore, $P(k + 1)$ is true.
 - Finally, by **strong induction**, every positive integer is a power of a prime or a product of powers of primes.

Mathematical Induction Summary

- A typical **proof by induction**, showing that $P(n)$ is true for all integers $n \geq b$, consists of three steps:
 - **Basis step:** prove $P(b)$ is true
 - **Inductive step:** prove one of the following
$$\forall k \geq b, P(k) \rightarrow P(k + 1) \text{ is true } \mathbf{OR}$$
$$\forall k \geq b, P(b) \wedge \dots \wedge P(k) \rightarrow P(k + 1) \text{ is true}$$
 - **Conclusion:** based on the (second) principle of mathematical induction, we conclude that $P(n)$ is true for all $n \geq b$.
- The assumption “ $P(k)$ is true” **OR** “ $P(1) \wedge P(2) \wedge \dots \wedge P(k)$ is true” is called the **inductive hypothesis (IH)**.
 - IH is used to prove “ $P(k + 1)$ is true”.

Recursion

Recursion

- **Recursion:** a method of solving a computational problem where its solution depends on solutions to **smaller instances of the same problem**.
- Recursive computer programs or algorithms often lead to **inductive analysis**.
- A classical **example of recursion** is the **Towers of Hanoi** Problem.

Towers of Hanoi

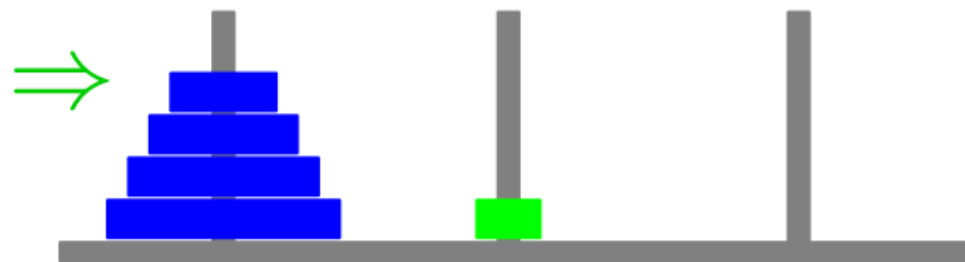
- **Problem:** Find an efficient way to move all of the disks from one peg to another.
 - 3 pegs and n disks of different sizes
 - A **legal move** takes a disk from one peg and moves it onto another peg so that it is **not on top of a smaller disk**.



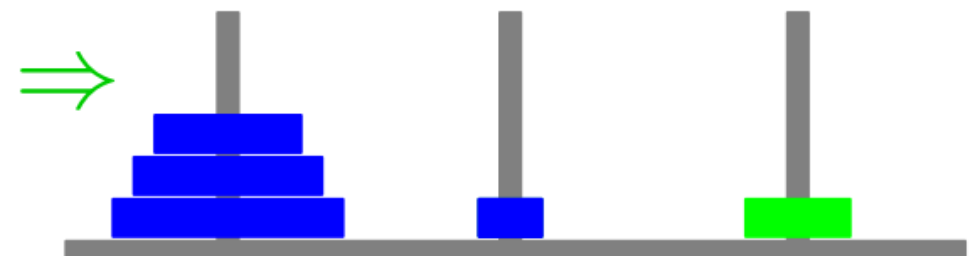
Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks from one peg to another.
 - 3 pegs and n disks of different sizes
 - A **legal move** takes a disk from one peg and moves it onto another peg so that it is **not on top of a smaller disk**.

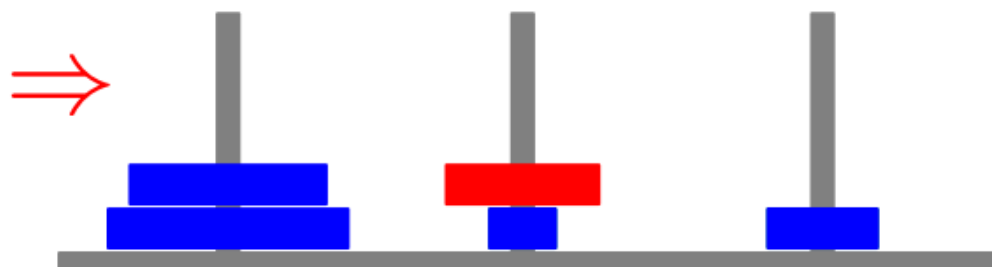
legal move



legal move



not legal

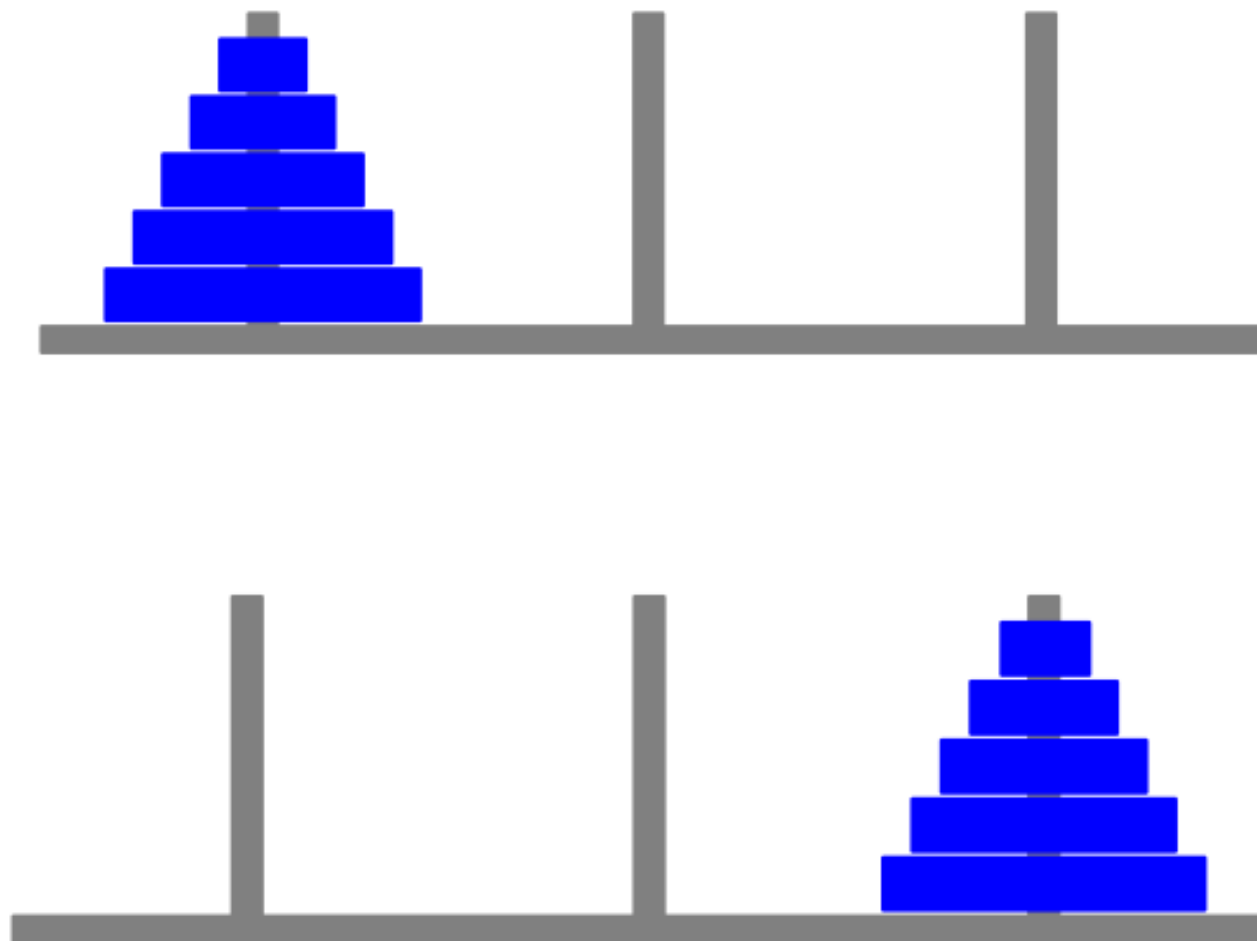


legal move



Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks from one peg to another, using only legal moves.



Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks **from one peg to another**, using **only legal moves**.
- **Solution** by recursion:
 - **Basis step:** If $n = 1$, moving one disk from one to another is easy.



- **Recursive step:** If $n > 1$, we need three steps:



Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks from one peg to another, using only legal moves.
- **Solution** by recursion:

```
3 public class Hanoi
4 {
5     move n disks from peg a to peg c using peg b
6     public void move(int n, char a, char b, char c)
7     {
8         if (n == 1)
9             System.out.println("plate " + n + " from " + a + " to " + c);
10        else
11        {
12            move(n-1, a, c, b);
13            System.out.println("plate " + n + " from " + a + " to " + c);
14            move(n-1, b, a, c);
15        }
16    }
17 }
18
```

Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks from one peg to another, using only legal moves.
- **Proof of correctness by induction:**
 - Let $P(n)$ be the proposition that the solution is correct for n .
 - **Basis step:** $P(1)$ is obviously true, i.e., the solution is correct when there is only one disk.
 - **Inductive step:** From the inductive hypothesis, i.e., $P(k)$ is true for an arbitrary positive integer k , we need to show that $P(k + 1)$ is true. That is, if our solution works for k disks, then we can build a correct solution for $k + 1$ disks, which is true by the recursive step:



- By mathematical induction, $P(n)$ is true for all positive integer n .

Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks from one peg to another, using only legal moves.
- **Solution** by recursion: *running time: # disk moves* $M(n) = ?$
 - **Basis step:** If $n = 1$, moving one disk from one to another is easy.



- **Recursive step:** If $n > 1$, we need three steps:



$$M(n) = 2M(n - 1) + 1 \text{ for } n > 1$$

Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks from one peg to another, using only legal moves.

- **Solving the running time:**

$$M(1) = 1 \qquad M(n) = 2M(n - 1) + 1 \text{ for } n > 1$$

- Iterating the above function gives:

$$M(1) = 1, M(2) = 3, M(3) = 7, M(4) = 15, M(5) = 31, \dots$$

- We can guess that $M(n) = 2^n - 1$ and prove it by induction:
Let $P(n)$ denote the above equation.

Basis step: $P(1)$ is true, because $M(1) = 1 = 2^1 - 1$.

Inductive step: Assume $P(k)$ is true for $k \geq 1$, i.e., $M(k) = 2^k - 1$.

Then $P(k + 1)$ is true: $M(k + 1) = 2M(k) + 1 = 2(2^k - 1) + 1 = 2^{k+1} - 1$

By mathematical induction, $P(n)$ is true for all positive n .

Towers of Hanoi

- **Problem:** Find an efficient way to move all of the disks from one peg to another, using only legal moves.
- Note that we applied induction twice:
 - first to prove **correctness** of the solution
 - second to derive the **closed-form running time** $M(n) = 2^n - 1$

Recurrences

Recurrences

- A **recurrence equation** or **recurrence** for a function defined on the set of integers $\geq b$ tells us how to **compute the n -th value from some or all the first $n - 1$ values**.
- To completely specify a function defined by a recurrence, we have to give the **initial condition(s)** (as known as the **base case(s)**) for the recurrence.
- Example: running time for Towers of Hanoi

$$M(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2M(n - 1) + 1 & \text{otherwise} \end{cases}$$

Example

- Let $S(n)$ be the number of subsets of a set of size n . We already learned that $S(n) = 2^n$, but now let us think about this recursively:

- Consider the 8 subsets of $\{1, 2, 3\}$:

\emptyset	$\{1\}$	$\{2\}$	$\{1, 2\}$
$\{3\}$	$\{1, 3\}$	$\{2, 3\}$	$\{1, 2, 3\}$

- The top 4 subsets are exactly the subsets of $\{1, 2\}$, while the bottom 4 subsets are the subsets of $\{1, 2\}$ with 3 added into each.
- So, we get a subset of $\{1, 2, 3\}$ either by taking a subset of $\{1, 2\}$ or by adding 3 to a subset of $\{1, 2\}$.
- This suggests that the recurrence for the number of subsets of an n -element set $\{1, 2, \dots, n\}$ is

$$S(n) = \begin{cases} 1 & \text{if } n = 0 \\ 2S(n-1) & \text{if } n \geq 1 \end{cases}$$

Example

- Let $S(n)$ be the number of subsets of a set of size n . We already learned that $S(n) = 2^n$, but now let us think about this recursively:
- Proof of correctness of the recurrence:

$$S(n) = \begin{cases} 1 & \text{if } n = 0 \\ 2S(n-1) & \text{if } n \geq 1 \end{cases}$$

- The subsets of $\{1, 2, \dots, n\}$ can be partitioned according to whether or not they contain the element n .

Each subset S containing n can be constructed in a unique fashion by adding n to the subset $S - \{n\}$ that does not contain n .

Each subset S not containing n can be constructed by removing n from the unique set $S \cup \{n\}$ that contains n .

So, the number of subsets containing n is exactly the same as the number of subsets not containing n .

- Therefore, if $n \geq 1$, then $S(n) = 2S(n-1)$.

Example

- Let $S(n)$ be the number of subsets of a set of size n . We already learned that $S(n) = 2^n$, but now let us think about this recursively:
- Proof of the closed form $S(n) = 2^n$ for recurrence:

$$S(n) = \begin{cases} 1 & \text{if } n = 0 \\ 2S(n-1) & \text{if } n \geq 1 \end{cases}$$

- *the proof (by induction) is left as an exercise*

Iterating a Recurrence

- Let $T(n) = rT(n - 1) + a$, where r and a are constants.
- Find a **recurrence** that expresses
 - $T(n)$ in terms of $T(n - 2)$
 - $T(n)$ in terms of $T(n - 3)$
 - $T(n)$ in terms of $T(n - 4)$
 - ...
- Can we generalize this to find a **closed-form solution**?

Iterating a Recurrence

◦ Note that $T(n) = rT(n-1) + a$ implies that

- $\forall i < n, T(n-i) = rT((n-i)-1) + a$

◦ Then, we have

$$\begin{aligned}T(n) &= rT(n-1) + a \\&= r(rT(n-2) + a) + a \\&= r^2T(n-2) + ra + a \\&= r^2(rT(n-3) + a) + ra + a \\&= r^3T(n-3) + r^2a + ra + a \\&= r^3(rT(n-4) + a) + r^2a + ra + a \\&= r^4T(n-4) + r^3a + r^2a + ra + a.\end{aligned}$$

◦ Guess: $T(n) = r^n T(0) + a \sum_{i=0}^{n-1} r^i$

Iterating a Recurrence

- The method we used to guess the solution is called **iterating the recurrence**, because we iteratively use the recurrence.
- Another approach is to iterate from the “**bottom-up**” instead of “**top-down**”.
- E.g., $T(n) = rT(n - 1) + a$, where r and a are constants.

$$T(0) = b$$

$$T(1) = rT(0) + a = rb + a$$

$$T(2) = rT(1) + a = r(rb + a) + a = r^2b + ra + a$$

$$T(3) = rT(2) + a = r^3b + r^2a + ra + a$$

- This would lead to the same guess: $T(n) = r^n T(0) + a \sum_{i=0}^{n-1} r^i$

Formula of Recurrences

- **Theorem:** If $T(n) = rT(n - 1) + a$, $T(0) = b$, and $r \neq 1$, then for all non-negative integers n , we have:

$$T(n) = r^n b + a \frac{1 - r^n}{1 - r}$$

- Proof by induction:

- **Basis step:** The formula is true for $n = 0$. *Why?*

$$T(0) = r^0 b + a \frac{1 - r^0}{1 - r} = b.$$

- **Inductive step:** ?

Formula of Recurrences

- **Theorem:** If $T(n) = rT(n - 1) + a$, $T(0) = b$, and $r \neq 1$, then for all non-negative integers n , we have:

$$T(n) = r^n b + a \frac{1 - r^n}{1 - r}$$

- Proof by induction:

- **Basis step:** The formula is true for $n = 0$: $T(0) = r^0 b + a \frac{1 - r^0}{1 - r} = b$.

- **Inductive step:**

$$\begin{aligned} T(n) &= rT(n - 1) + a \\ &= r \left(r^{n-1} b + a \frac{1 - r^{n-1}}{1 - r} \right) + a \\ &= r^n b + \frac{ar - ar^n}{1 - r} + a \\ &= r^n b + \frac{ar - ar^n + a - ar}{1 - r} \\ &= r^n b + a \frac{1 - r^n}{1 - r}. \end{aligned}$$

Formula of Recurrences

- **Theorem:** If $T(n) = rT(n - 1) + a$, $T(0) = b$, and $r \neq 1$, then for all non-negative integers n , we have:

$$T(n) = r^n b + a \frac{1 - r^n}{1 - r}$$

- Example: $T(n) = 3T(n - 1) + 2$, $T(0) = 5$
 - Plugging $r = 3$, $a = 2$, $b = 5$ in the formula, we have:

$$T(n) = 3^n \cdot 5 + 2 \frac{1 - 3^n}{1 - 3} = 3^n \cdot 6 - 1$$

First-Order Linear Recurrences

- A recurrence of the form $T(n) = f(n)T(n - 1) + g(n)$ is called a **first-order linear** recurrence.
 - **First order:** $T(n)$ depends upon going back one step, i.e., $T(n - 1)$
e.g., $T(n) = T(n - 1) + 2T(n - 2)$ is a **second-order** recurrence
 - **Linear:** the $T(n - 1)$ only appears to the **first power**.
e.g., $T(n) = (T(n - 1))^2 + 3$ is a **non-linear** first-order recurrence
- When $f(n)$ is a **constant**, say r , the general solution is almost as easy as we derived before. Iterating the recurrence gives

$$\begin{aligned} T(n) &= rT(n - 1) + g(n) \\ &= r(rT(n - 2) + g(n - 1)) + g(n) \\ &= r^2 T(n - 2) + rg(n - 1) + g(n) \\ &\vdots \\ &= r^n T(0) + \sum_{i=0}^{n-1} r^i g(n - i) \end{aligned}$$

First-Order Linear Recurrences

- **Theorem:** For any positive constants a and r , and any function g defined on nonnegative integers, the **solution to the first-order linear recurrence**

$$T(n) = \begin{cases} rT(n-1) + g(n) & \text{if } n > 0 \\ a & \text{if } n = 0 \end{cases}$$

is

$$T(n) = r^n a + \sum_{i=1}^n r^{n-i} g(i).$$

Exercise (5 mins)

- Solve $T(n) = 4T(n - 1) + 2^n$ ($n > 0$) with $T(0) = 6$

Hint: express $T(n)$ in terms of 4^n and 2^n

- **Theorem:** For any positive constants a and r , and any function g defined on nonnegative integers, the solution to the first-order linear recurrence

$$T(n) = \begin{cases} rT(n-1) + g(n) & \text{if } n > 0 \\ a & \text{if } n = 0 \end{cases}$$

is

$$T(n) = r^n a + \sum_{i=1}^n r^{n-i} g(i).$$

Divide-and-Conquer Recurrences

Divide and Conquer

- A **divide-and-conquer** algorithm **recursively breaks down a problem into two or more sub-problems of the same or related type**, until these become simple enough to be solved directly. The solutions to the sub-problems are then combined to give a solution to the original problem.
- Divide-and-conquer recurrences are usually of the form:

$$T(n) = \begin{cases} \text{something given} & \text{if } n \leq n_0 \\ r \cdot T(n/m) + g(n) & \text{if } n > n_0 \end{cases}$$

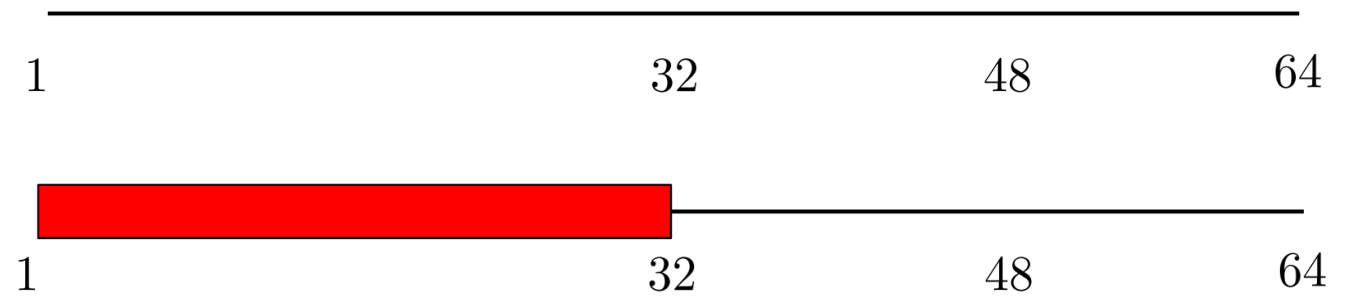
Binary Search

- Someone has chosen a number x between 1 and n . We need to discover x .
- We only need to ask two types of questions:
 - Is x greater than k ?
 - Is x equal to k ?
- Our strategy will be to always ask greater than questions, at each step halving our search range, until the range only contains one number, when we ask a final equal to question.

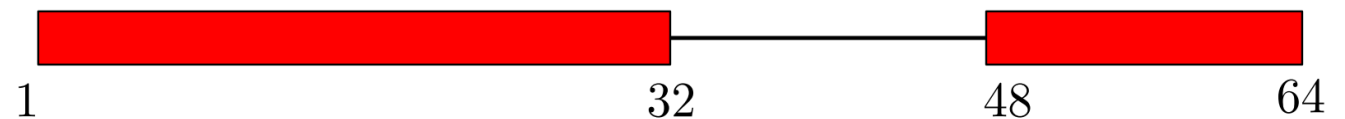
Binary Search

○ Example: $n = 64$, $x = 35$

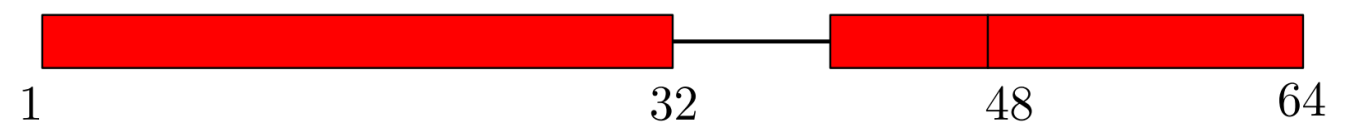
$x > 32?$ Yes



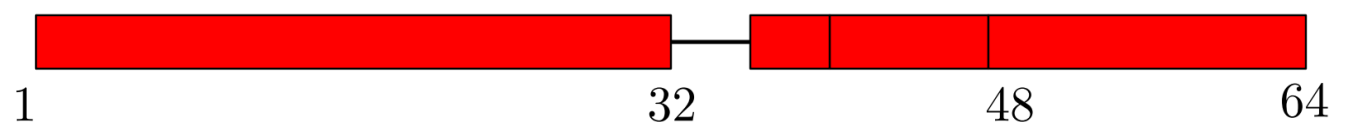
$x > 48?$ No



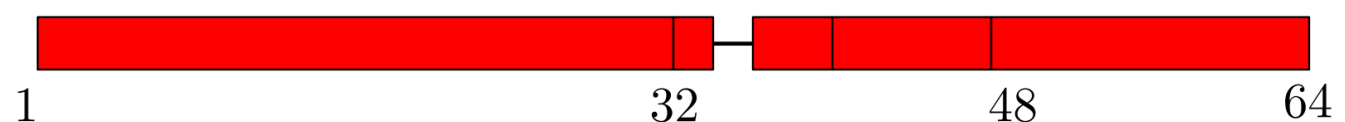
$x > 40?$ No



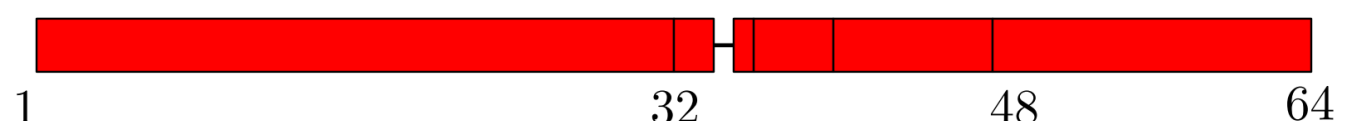
$x > 36?$ No



$x > 34?$ Yes



$x > 35?$ No



$x = 35?$ Yes



Binary Search

- Method: Each guess **reduces** the problem to one in which **the range is only half** as big.
- This **divides** the original problem into one that is only half as big; we can now **(recursively) conquer** this smaller problem.
- When n is a power of 2, $T(n)$, the number of comparisons in a binary search on $[1, n]$, satisfies

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + 1 & \text{if } n \geq 2 \end{cases}$$

- The inductive correctness proof is similar to the tower of Hanoi:
 - **Basis Step** ($n = 1$): only **one “equal to”** comparison is needed
 - **Inductive Step** ($n > 1$): **one “great than”** comparison + time to perform **binary search on the remaining $n/2$ terms**

Iterating Recurrences

- Example 1:

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + 1 & \text{if } n \geq 2 \end{cases}$$

- Solve it by algebraically iterating the recurrence:

$$\begin{aligned} T(n) &= T\left(\frac{n}{2}\right) + 1 &= \left(T\left(\frac{n}{2^2}\right) + 1\right) + 1 \\ &= T\left(\frac{n}{2^2}\right) + 2 &= \left(T\left(\frac{n}{2^3}\right) + 1\right) + 2 \\ &= T\left(\frac{n}{2^3}\right) + 3 \end{aligned}$$

$$\begin{aligned} &\vdots \\ &= T\left(\frac{n}{2^i}\right) + i \end{aligned}$$

End when $i = \log_2 n$

$$\begin{aligned} &\vdots \\ &= T\left(\frac{n}{2^{\log_2 n}}\right) + \log_2 n = 1 + \log_2 n \end{aligned}$$

Iterating Recurrences

- Example 1:

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + 1 & \text{if } n \geq 2 \end{cases}$$

- We just showed by algebraically **iterating the recurrence** that the solution is $T(n) = 1 + \log_2 n$.
- **Note:** Technically, we still need to use **induction** to prove that our solution is correct. Practically, we **never** explicitly perform this step, since it is obvious how the induction would work.

Iterating Recurrences

- Example 2:

$$T(n) = \begin{cases} T(1) & \text{if } n = 1 \\ 2T(n/2) + n & \text{if } n \geq 2 \end{cases}$$

- This corresponds to solving a problem of size n , by
 - (i) solving 2 subproblems of size $n/2$
 - (ii) doing n units of additional workor using $T(1)$ work for “bottom” case of $n = 1$
- This is exactly how merge sort (from an algorithm course) works.
- How to solve it by algebraically iterating the recurrence?

Iterating Recurrences

- Example 2:

$$T(n) = \begin{cases} T(1) & \text{if } n = 1 \\ 2T(n/2) + n & \text{if } n \geq 2 \end{cases}$$

- Solve it by algebraically iterating the recurrence:

$$T(n) = 2T\left(\frac{n}{2}\right) + n = 2\left(2T\left(\frac{n}{4}\right) + \frac{n}{2}\right) + n$$

$$= 4T\left(\frac{n}{4}\right) + 2n = 4\left(2T\left(\frac{n}{8}\right) + \frac{n}{4}\right) + 2n$$

$$= 8T\left(\frac{n}{8}\right) + 3n$$

$$\vdots \quad \vdots$$
$$= 2^i T\left(\frac{n}{2^i}\right) + in$$

$$\vdots \quad \vdots$$
$$= 2^{\log_2 n} T\left(\frac{n}{2^{\log_2 n}}\right) + (\log_2 n)n = nT(1) + n \log_2 n$$

Iterating Recurrences

- Example 3:

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + n & \text{if } n \geq 2 \end{cases}$$

- Solve it by algebraically iterating the recurrence:

$$\begin{aligned} T(n) &= T\left(\frac{n}{2}\right) + n \\ &= T\left(\frac{n}{2^2}\right) + \frac{n}{2} + n \\ &= T\left(\frac{n}{2^3}\right) + \frac{n}{2^2} + \frac{n}{2} + n \\ &\quad \vdots \\ &= T\left(\frac{n}{2^i}\right) + \frac{n}{2^{i-1}} + \cdots + \frac{n}{2^2} + \frac{n}{2} + n \\ &\quad \vdots \\ &= T\left(\frac{n}{2^{\log_2 n}}\right) + \frac{n}{2^{\log_2 n - 1}} + \cdots + \frac{n}{2^2} + \frac{n}{2} + n \\ &= 1 + 2 + 2^2 + \cdots + \frac{n}{2^2} + \frac{n}{2} + n = 2n - 1 \end{aligned}$$

Exercise (5 mins)

- Solve this recurrence by algebraically iterating the recurrence:

$$T(n) = \begin{cases} 1 & \text{if } n < 3 \\ 3T(n/3) + n & \text{if } n \geq 3 \end{cases}$$

- Example 3:

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + n & \text{if } n \geq 2 \end{cases}$$

- Solve it by algebraically iterating the recurrence:

$$\begin{aligned} T(n) &= T\left(\frac{n}{2}\right) + n \\ &= T\left(\frac{n}{2^2}\right) + \frac{n}{2} + n \\ &= T\left(\frac{n}{2^3}\right) + \frac{n}{2^2} + \frac{n}{2} + n \\ &\quad \vdots \\ &= T\left(\frac{n}{2^i}\right) + \frac{n}{2^{i-1}} + \cdots + \frac{n}{2^2} + \frac{n}{2} + n \\ &\quad \vdots \\ &= T\left(\frac{n}{2^{\log_2 n}}\right) + \frac{n}{2^{\log_2 n - 1}} + \cdots + \frac{n}{2^2} + \frac{n}{2} + n \\ &= 1 + 2 + 2^2 + \cdots + \frac{n}{2^2} + \frac{n}{2} + n \end{aligned}$$

Iterating Recurrences

- Example 4:

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 4T(n/2) + n & \text{if } n \geq 2 \end{cases}$$

- Solve it by algebraically iterating the recurrence:

$$\begin{aligned} T(n) &= 4T\left(\frac{n}{2}\right) + n &&= 4\left(4T\left(\frac{n}{2^2}\right) + \frac{n}{2}\right) + n \\ &= 4^2 T\left(\frac{n}{2^2}\right) + \frac{4}{2}n + n &&= 4^2\left(4T\left(\frac{n}{2^3}\right) + \frac{n}{2^2}\right) + \frac{4}{2}n + n \\ &= 4^3 T\left(\frac{n}{2^3}\right) + \frac{4^2}{2^2}n + \frac{4}{2}n + n \\ &\vdots \\ &= 4^i T\left(\frac{n}{2^i}\right) + \frac{4^{i-1}}{2^{i-1}}n + \cdots + \frac{4^2}{2^2}n + n \\ &\vdots \\ &= 4^{\log_2 n} T\left(\frac{n}{2^{\log_2 n}}\right) + \frac{4^{\log_2 n - 1}}{2^{\log_2 n - 1}}n + \cdots + \frac{4}{2}n + n = 2n^2 - n \end{aligned}$$

Three Different Behaviors

- Compare the iteration for the following recurrences ($T(1) = 1$):
 - $T(n) = T(n/2) + n$ $T(n) = 2n - 1 = \Theta(n)$
 - $T(n) = 2T(n/2) + n$ $T(n) = n + n \log_2 n = \Theta(n \log n)$
 - $T(n) = 4T(n/2) + n$ $T(n) = 2n^2 - n = \Theta(n^2)$
 - All three recurrences iterate $\log_2 n$ times. The size of subproblem in next iteration is **half** the size in the preceding iteration level.
- **Theorem:** Suppose that we have a recurrence $T(n) = aT(n/2) + n$, where $a \geq 1$ and $T(1) = \Theta(1)$. Then we have the following big Θ bounds on the solution:
 - If $a < 2$, then $T(n) = \Theta(n)$. * *the proof is left as an exercise*
 - If $a = 2$, then $T(n) = \Theta(n \log n)$. * *already proved in Example 2*
 - If $a > 2$, then $T(n) = \Theta(n^{\log_2 a})$. * *now let us prove this*

Three Different Behaviors

- Let $n = 2^i$. Prove that if $a > 2$, then $T(n) = \Theta(n^{\log_2 a})$.
- Iterating $T(n) = aT(n/2) + n$ as in Example 4 gives:

$$T(n) = a^i T\left(\frac{n}{2^i}\right) + \left(\frac{a^{i-1}}{2^{i-1}} + \frac{a^{i-2}}{2^{i-2}} + \cdots + \frac{a}{2} + 1\right) n$$

$$T(n) = a^{\log_2 n} T(1) + n \sum_{i=0}^{\log_2 n - 1} \left(\frac{a}{2}\right)^i$$

Work at
“bottom”

Iterated
Work

- Since $a > 2$, the geometric series is Θ of the largest term:

$$n \sum_{i=0}^{\log_2 n - 1} \left(\frac{a}{2}\right)^i = n \frac{1 - (a/2)^{\log_2 n}}{1 - a/2} = n\Theta\left(\left(\frac{a}{2}\right)^{\log_2 n - 1}\right)$$

Three Different Behaviors

- Let $n = 2^i$. Prove the third case: If $a > 2$, then $T(n) = \Theta(n^{\log_2 a})$.
- Iterating $T(n) = aT(n/2) + n$ as in Example 4 gives:

$$T(n) = a^{\log_2 n} T(1) + n \sum_{i=0}^{\log_2 n - 1} \left(\frac{a}{2}\right)^i$$

- Since $a > 2$, the geometric series is Θ of the largest term:

$$n \sum_{i=0}^{\log_2 n - 1} \left(\frac{a}{2}\right)^i = n \frac{1 - (a/2)^{\log_2 n}}{1 - a/2} = n\Theta\left(\left(\frac{a}{2}\right)^{\log_2 n - 1}\right)$$

$$n \left(\frac{a}{2}\right)^{\log_2 n - 1} = \frac{2}{a} \cdot \frac{n \cdot a^{\log_2 n}}{2^{\log_2 n}} = \frac{2}{a} \cdot \frac{n \cdot a^{\log_2 n}}{n} = \frac{2}{a} \cdot a^{\log_2 n}$$

$$a^{\log_2 n} = \left(2^{\log_2 a}\right)^{\log_2 n} = \left(2^{\log_2 n}\right)^{\log_2 a} = n^{\log_2 a}$$

$$a^{\log_2 n} T(1) + n \sum_{i=0}^{\log_2 n - 1} \left(\frac{a}{2}\right)^i = \Theta\left(n^{\log_2 a}\right)$$

Three Different Behaviors

- **Theorem:** Suppose that we have a recurrence $T(n) = aT(n/2) + n$, where $a \geq 1$ and $T(1) = \Theta(1)$. Then we have the following big Θ bounds on the solution:
 - If $a < 2$, then $T(n) = \Theta(n)$. ** proof is left as an exercise*
 - If $a = 2$, then $T(n) = \Theta(n \log n)$. ** already proved in Example 2*
 - If $a > 2$, then $T(n) = \Theta(n^{\log_2 a})$. ** just proved*
- **Master Theorem:** Consider a recurrence $T(n) = aT(n/b) + cn^d$, where $a \geq 1$, $c > 0$, $d \geq 0$, integer $b \geq 2$, and $T(1) = \Theta(1)$. Then we have the following big Θ bounds on the solution:
 - If $a < b^d$, then $T(n) = \Theta(n^d)$.
 - If $a = b^d$, then $T(n) = \Theta(n^d \log n)$.
 - If $a > b^d$, then $T(n) = \Theta(n^{\log_b a})$.

07 Counting

To be continued...