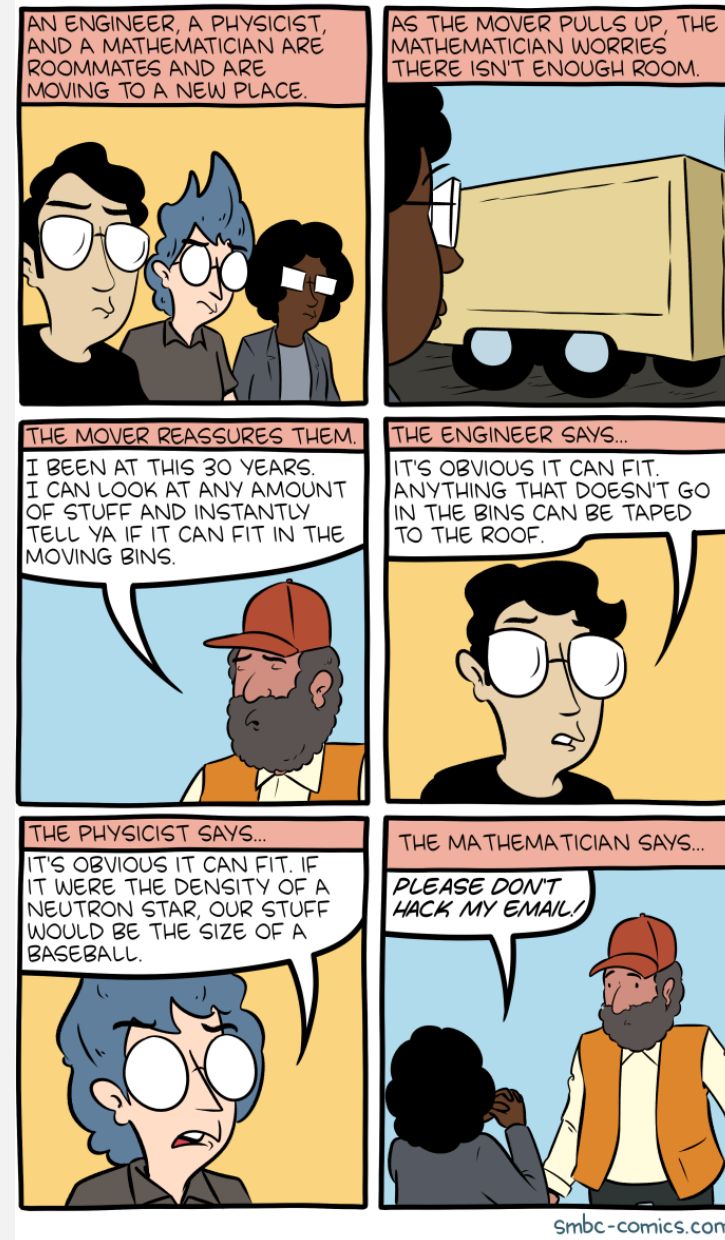


CS420, Ch7 Time Complexity

Wed April 14, 2021

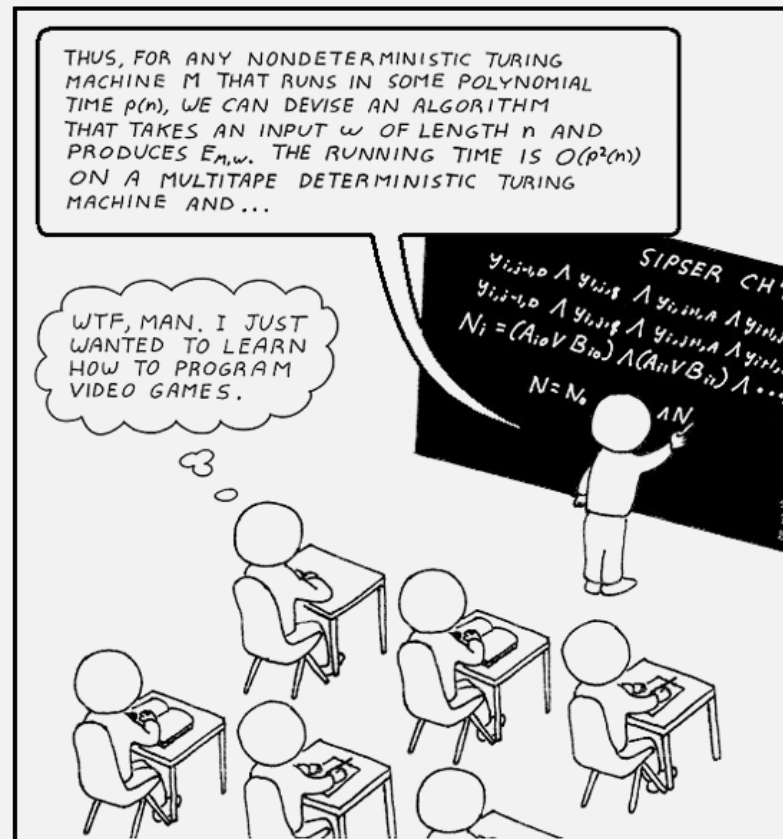


Announcements

- Reminder: No class next Monday 4/19!
- HW9
 - Due date extended to Tues 4/20 11:59pm EST
- HW10 coming soon!
- FAQ: How many HWs left?
 - Total: 12 HWs
- FAQ: What's my grade?
 - All your scores are visible in Gradescope
 - Letter grade brackets: 90s -> A, 80s, -> B, etc.
 - See: CS420 Spring 2021 Course Page -> Course Policies -> Grading
 - Final grade, incl. dropped hw, particip, not calculated until end of semester



Flashback: Single-tape TM “equiv to” Nondet. TM



Flashback: Single-tape TM “equiv to” Nondet. TM

- Deterministic TM simulating nondeterministic TM:

1. Number the nodes at each step
2. Deterministically check every path, in breadth-first order (restart at top each time)

- 1
- 1-1
- 1-2
- 1-1-1
- 1-1-2
- and so on

“This is the most inefficient algorithm ever”
--- CS420 Spring 2021 student

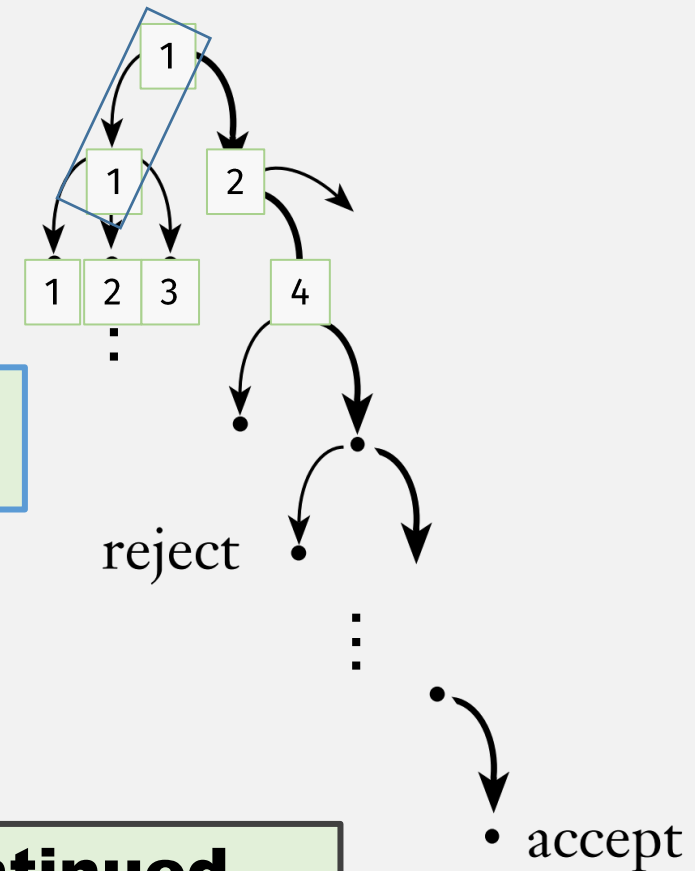
Exactly how inefficient is it???

Now we’ll start to count “# of steps”

3. Accept if accepting config found

To be continued ...

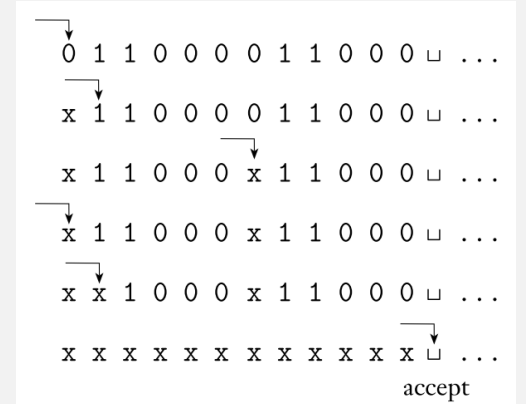
Nondeterministic computation



Simpler Example: $A = \{0^k 1^k \mid k \geq 0\}$

$M_1 =$ “On input string w :

1. Scan across the tape and *reject* if a 0 is found to the right of a 1.
2. Repeat if both 0s and 1s remain on the tape:
3. Scan across the tape, crossing off a single 0 and a single 1.
4. If 0s still remain after all the 1s have been crossed off, or if 1s still remain after all the 0s have been crossed off, *reject*. Otherwise, if neither 0s nor 1s remain on the tape, *accept*.”



Number of steps (worst case), $n =$ length of input:

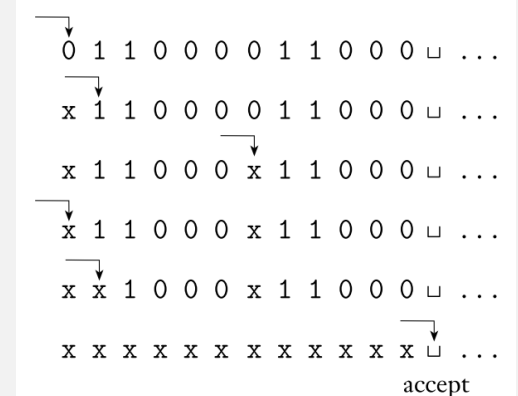
➤ TM Line 1:

- n steps to scan + n steps to return to beginning = $2n$ steps

Simpler Example: $A = \{0^k 1^k \mid k \geq 0\}$

$M_1 =$ “On input string w :

1. Scan across the tape and *reject* if a 0 is found to the right of a 1.
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Number of steps (worst case), $n =$ length of input:

• TM Line 1:

- n steps to scan + n steps to return to beginning = $2n$ steps

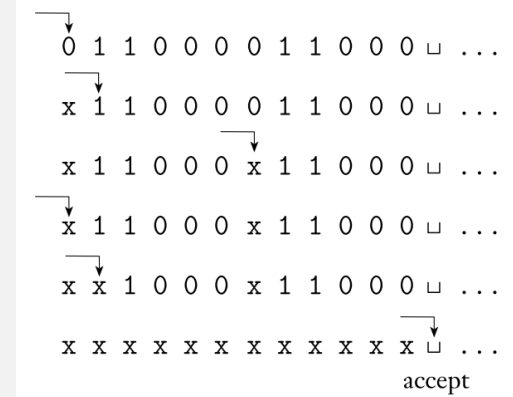
➤ Lines 2 and 3 (loop):

- Each iter: $n/2$ steps to find “1” + $n/2$ steps to return = n steps
- Num iters: Each scan crosses off 2 chars, so at most $n/2$ scans
- Total = each iter times num iters = $n (n/2) =$ $n^2/2$ steps

Simpler Example: $A = \{0^k 1^k \mid k \geq 0\}$

$M_1 =$ “On input string w :

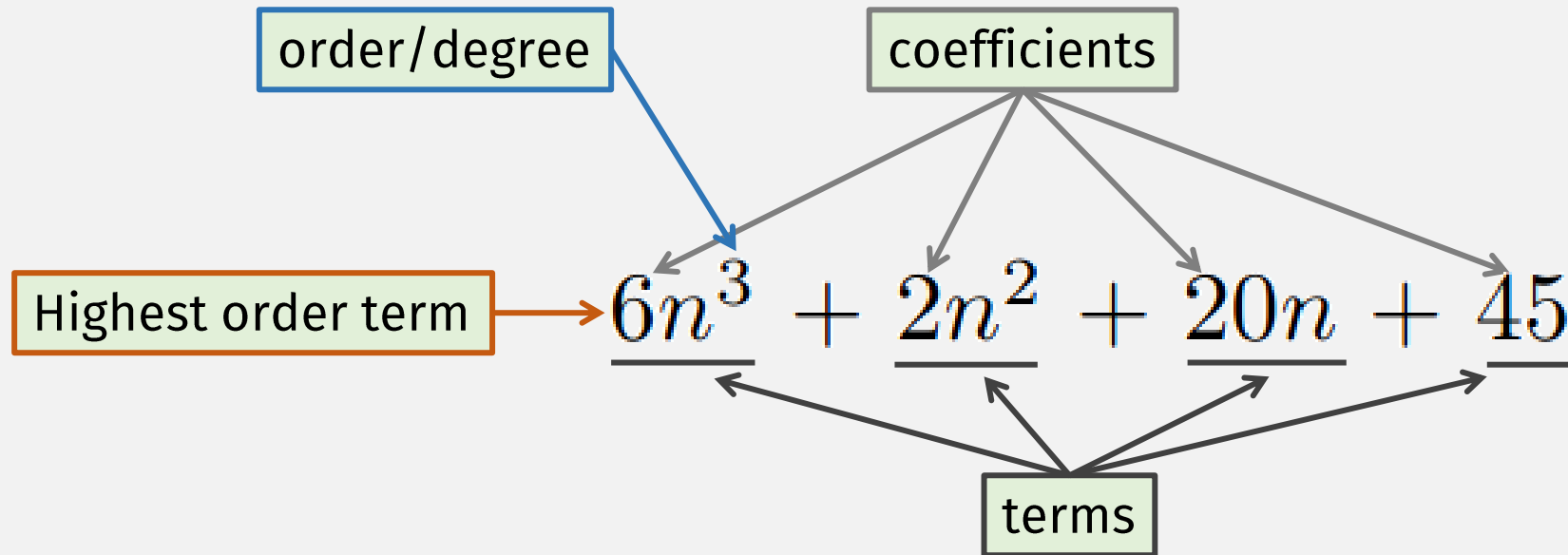
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Number of steps (worst case), $n =$ length of input:

- TM Line 1:
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 - Num iters: Each scan crosses off 2 chars, so at most $n/2$ scans
 - Total = each iter times num iters = $n (n/2) =$ $n^2/2$ steps
- Line 4:
 - n steps to scan input one more time
- Total: $2n + n^2/2 + n =$ $n^2/2 + 3n$ steps

Interlude: Polynomials



Definition: Time Complexity

NOTE: n has no units, it's only roughly "length" of the input

DEFINITION 7.1

But n can be not only #characters, but also #states, #nodes, etc.

Let M be a deterministic Turing machine that halts on all inputs. The *running time* or *time complexity* of M is the function $f: \mathcal{N} \rightarrow \mathcal{N}$, where $f(n)$ is the maximum number of steps that M uses on any input of length n . If $f(n)$ is the running time of M , we say that M runs in time $f(n)$ and that M is an $f(n)$ time Turing machine. Customarily we use n to represent the length of the input.

We can use any of things for n , bc they're correlated with input length

- Machine M_1 that decides $A = \{0^k 1^k \mid k \geq 0\}$
 - Running Time: $n^2/2 + 3n$

$M_1 =$ "On input string w :

1. Scan across the tape and *reject* if a 0 is found to the right of a 1.
2. Repeat if both 0s and 1s remain on the tape:
3. Scan across the tape, crossing off a single 0 and a single 1.
4. If 0s still remain after all the 1s have been crossed off, or if 1s still remain after all the 0s have been crossed off, *reject*. Otherwise, if neither 0s nor 1s remain on the tape, *accept*."

Interlude: Asymptotic Analysis

- Total: $n^2 + 3n$
 - If $n = 1$
 - $n^2 = 1$
 - $3n = 3$
 - Total = 4
 - If $n = 10$
 - $n^2 = 100$
 - $3n = 30$
 - Total = 130
 - If $n = 100$
 - $n^2 = 10000$
 - $3n = 300$
 - Total = 10300
 - If $n = 1000$
 - $n^2 = 1000000$
 - $3n = 3000$
 - Total = 1003000
- $n^2 + 3n \approx n^2$ as n gets large
- asymptotic analysis only cares about large n

Definition: Big- O Notation

DEFINITION 7.2

Let f and g be functions $f, g: \mathcal{N} \rightarrow \mathcal{R}^+$. Say that $f(n) = O(g(n))$ if positive integers c and n_0 exist such that for every integer $n \geq n_0$,

$$f(n) \leq c g(n).$$

When $f(n) = O(g(n))$, we say that $g(n)$ is an *upper bound* for $f(n)$, or more precisely, that $g(n)$ is an *asymptotic upper bound* for $f(n)$, to emphasize that we are suppressing constant factors.

- In English: Keep only highest order term, drop all coefficients
- Machine M_1 that decides $A = \{0^k 1^k \mid k \geq 0\}$
 - Is an $n^2 + 3n$ time Turing machine
 - Is an $O(n^2)$ time Turing machine
 - Has asymptotic upper bound $O(n^2)$

Definition: Small- o Notation (less used)

DEFINITION 7.5

Let f and g be functions $f, g: \mathcal{N} \rightarrow \mathcal{R}^+$. Say that $f(n) = o(g(n))$ if

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0.$$

In other words, $f(n) = o(g(n))$ means that for any real number $c > 0$, a number n_0 exists, where $f(n) < c g(n)$ for all $n \geq n_0$.

- Analogy:
 - Big- O : \leq :: small- o : $<$

DEFINITION 7.2

Let f and g be functions $f, g: \mathcal{N} \rightarrow \mathcal{R}^+$. Say that $f(n) = O(g(n))$ if positive integers c and n_0 exist such that for every integer $n \geq n_0$,

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When $f(n) = O(g(n))$, we say that $g(n)$ is an **upper bound** for $f(n)$, or more precisely, that $g(n)$ is an **asymptotic upper bound** for $f(n)$, to emphasize that we are suppressing constant factors.

Big- O arithmetic

- $O(\mathbf{n}^2) + O(\mathbf{n}^2)$
= $O(\mathbf{n}^2)$

- $O(\mathbf{n}^2) + O(\mathbf{n})$
= $O(\mathbf{n}^2)$

Definition: Time Complexity Classes

DEFINITION 7.7

Let $t: \mathcal{N} \rightarrow \mathcal{R}^+$ be a function. Define the *time complexity class*, $\text{TIME}(t(n))$, to be the collection of all languages that are decidable by an $O(t(n))$ time Turing machine.

TMs have a running time,
languages have a complexity class

- Machine M_1 that decides $A = \{0^k 1^k \mid k \geq 0\}$
 - Is an $O(\mathbf{n}^2)$ running time Turing machine
 - So A is in $\text{TIME}(\mathbf{n}^2)$

A Faster Machine? $A = \{0^k 1^k \mid k \geq 0\}$

M_2 = “On input string w :

1. Scan across the tape and *reject* if a 0 is found to the right of a 1.
2. Repeat as long as some 0s and some 1s remain on the tape:
3. Scan across the tape, checking whether the total number of 0s and 1s remaining is even or odd. If it is odd, *reject*.
4. Scan again across the tape, crossing off every other 0 starting with the first 0, and then crossing off every other 1 starting with the first 1.
5. If no 0s and no 1s remain on the tape, *accept*. Otherwise, *reject*.”

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Number of steps (worst case), n = length of input:

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Number of steps (worst case), $n =$ length of input:

➤ Line 1:

- n steps to scan + n steps to return to beginning = $O(n)$ steps

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Number of steps (worst case), $n =$ length of input:

• Line 1:

- n steps to scan + n steps to return to beginning = $O(n)$ steps

➤ Lines 2, 3, 4 (loop):

- Each iter: a scan takes $O(n)$ steps
- Num iters: Each iter crosses off *half* the chars, so at most $O(\log n)$ scans
- Total: $O(n) * O(\log n) =$ $O(n \log n)$ steps

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- Line 5:
 - $O(n)$ steps to scan input one more time

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Number of steps (worst case), n = length of input:

- Line 1:
 - n steps to scan + n steps to return to beginning = $O(n)$ steps
- Lines 2, 3, 4 (loop):
 - Each iter: a scan takes $O(n)$ steps
 - Num iters: Each iter crosses off half the chars, so at most $O(\log n)$ scans
 - Total: $O(n) * O(\log n) = \underline{O(n \log n)}$ steps
- Line 5:
 - $O(n)$ steps to scan input one more time
- Total: $O(n) + O(n \log n) + O(n) = \underline{O(n \log n)}$ steps

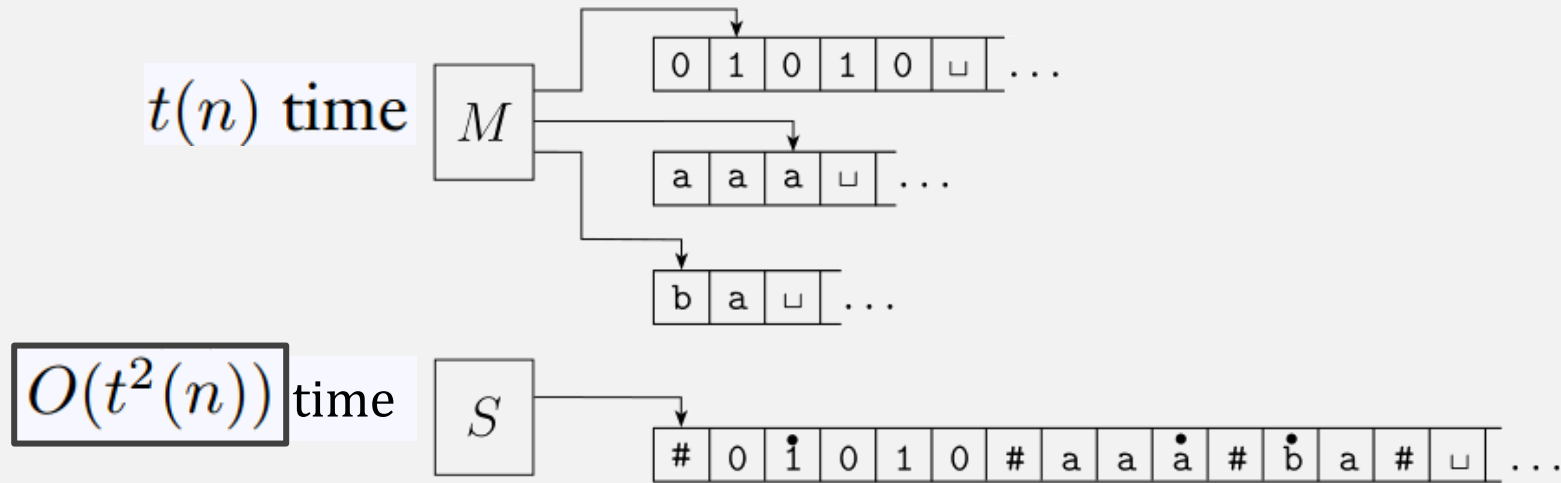
Interlude: Logarithms

- $2^x = y$
- $\log_2 y = x$
- $\log_2 n = O(\mathbf{\log n})$
 - “divide and conquer” algorithms = $O(\mathbf{\log n})$
 - E.g., binary search
- In computer science, **base-2 is the only base!**

Terminology: Categories of Bounds

- Exponential time
 - $O(2^{n^c})$, for $c > 0$ (always base 2)
- Polynomial time
 - $O(n^c)$, for $c > 0$
- Quadratic time (special case of polynomial time)
 - $O(n^2)$
- Linear time (special case of polynomial time)
 - $O(n)$
- Log time
 - $O(\log n)$

Multi-tape vs Single-tape TMs: # of Steps



- For single-tape TM to simulate 1 step of multi-tape:
 - Scan to find all “heads” = $O(\text{length of all } M\text{'s tapes})$
 - “Execute” transition at all the heads = $O(\text{length of all } M\text{'s tapes})$
- # single-tape steps to simulate 1 multitape step (worst case)
 - = $O(\text{length of all } M\text{'s tapes})$
 - = $O(t(n))$ (If M spends all its steps expanding its tapes)
- Total steps (single tape): $O(t(n))$ per step $\times t(n)$ steps = $O(t^2(n))$

Single-tape TM vs Nondet. TM: # of steps

- Deterministic TM simulating nondeterministic TM:
 - Number the nodes at each step
 - Deterministically check every path, in breadth-first order (restart at top each time)
 - 1
 - 1-1
 - 1-2
 - 1-1-1
 - 1-1-2
 - and so on
 - Accept if accepting config found

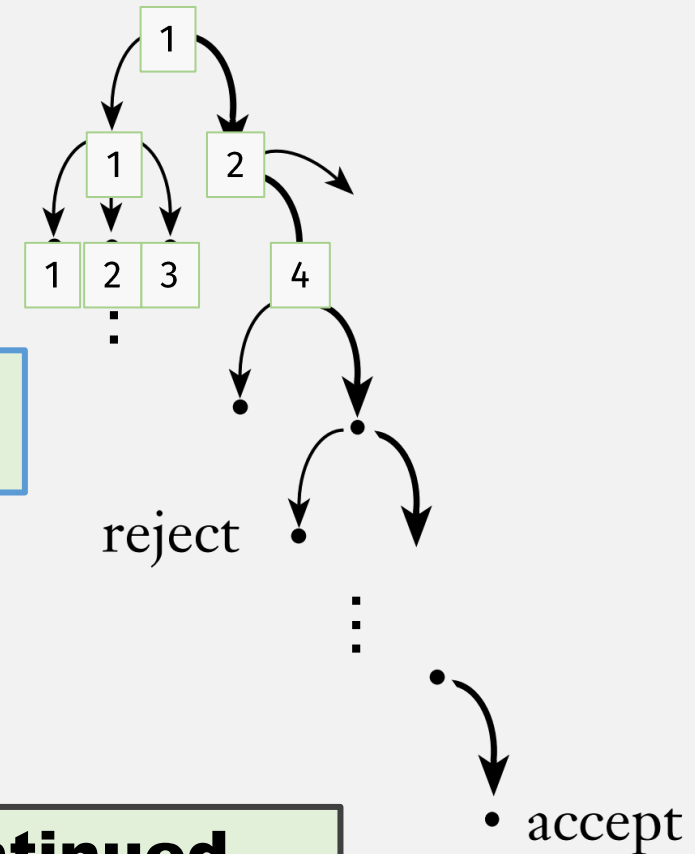
“This is the most inefficient algorithm ever”
--- CS420 Spring 2021 student

Exactly how inefficient is it???

Now we'll start to count “# of steps”

To be continued ...

Nondeterministic computation

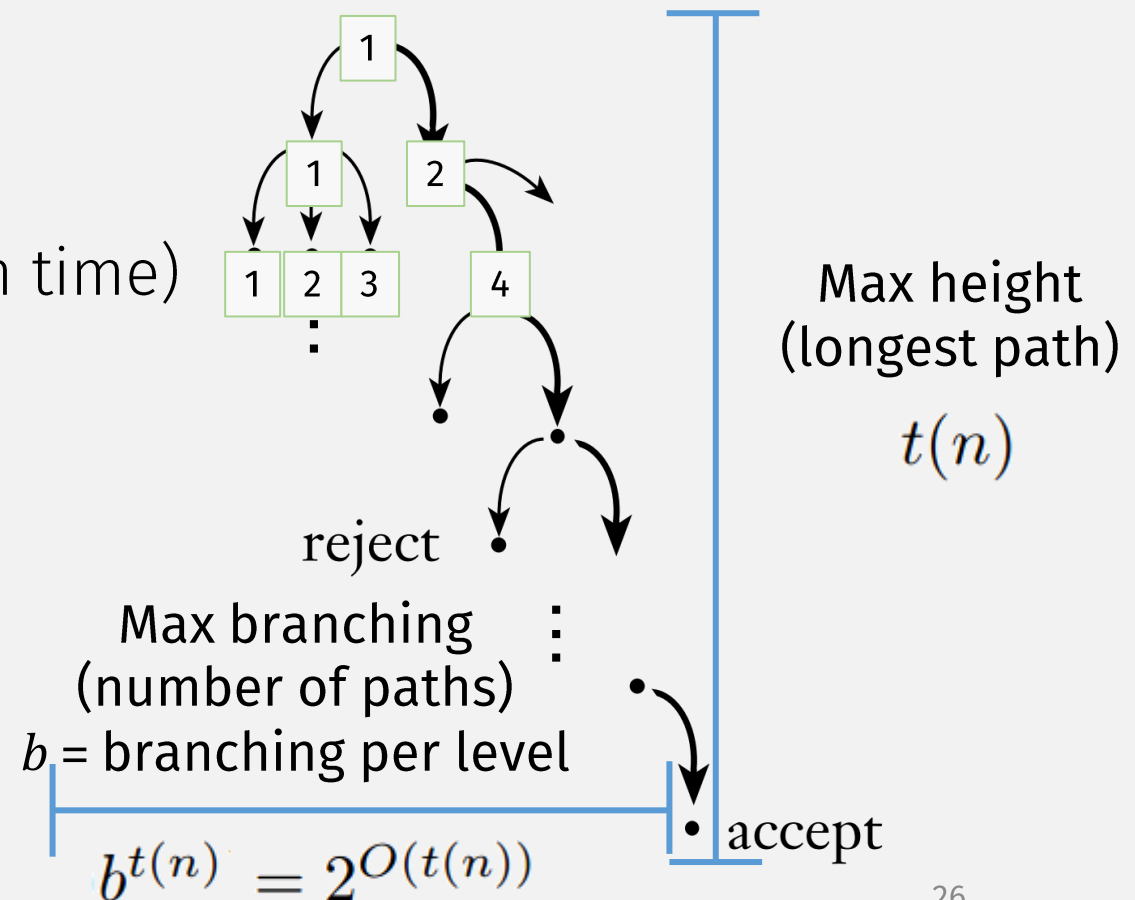


Single-tape TM vs Nondet. TM: # of steps

$2^{O(t(n))}$ time

- Deterministic TM simulating nondeterministic TM: $\leftarrow t(n)$ time
 - Number the nodes at each step
 - Deterministically check every path, in breadth-first order (restart at top each time)
 - 1
 - 1-1
 - 1-2
 - 1-1-1
 - 1-1-2
 - and so on
 - Accept if accepting config found

Nondeterministic computation



Summary

- If multi-tape TM: $t(n)$ time
- Then equivalent single-tape TM: $O(t^2(n))$
 - Quadratically slower
- If non-deterministic TM: $t(n)$ time
- Then equivalent single-tape TM: $2^{O(t(n))}$
 - Exponentially slower

Next time: Specific Complexity Classes

DEFINITION 7.12

P is the class of languages that are decidable in polynomial time on a deterministic single-tape Turing machine. In other words,

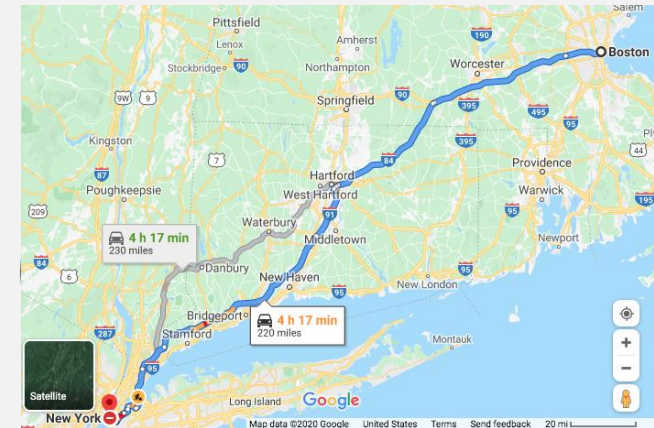
$$P = \bigcup_k \text{TIME}(n^k).$$

- Corresponds to “realistically” solvable problems
- In this class:
 - Problems in P = “solvable”
 - Problems outside P = “unsolvable”
 - These are usually “brute force” solutions that “try all possible inputs”

Next time: A Graph Theorem: $PATH \in P$

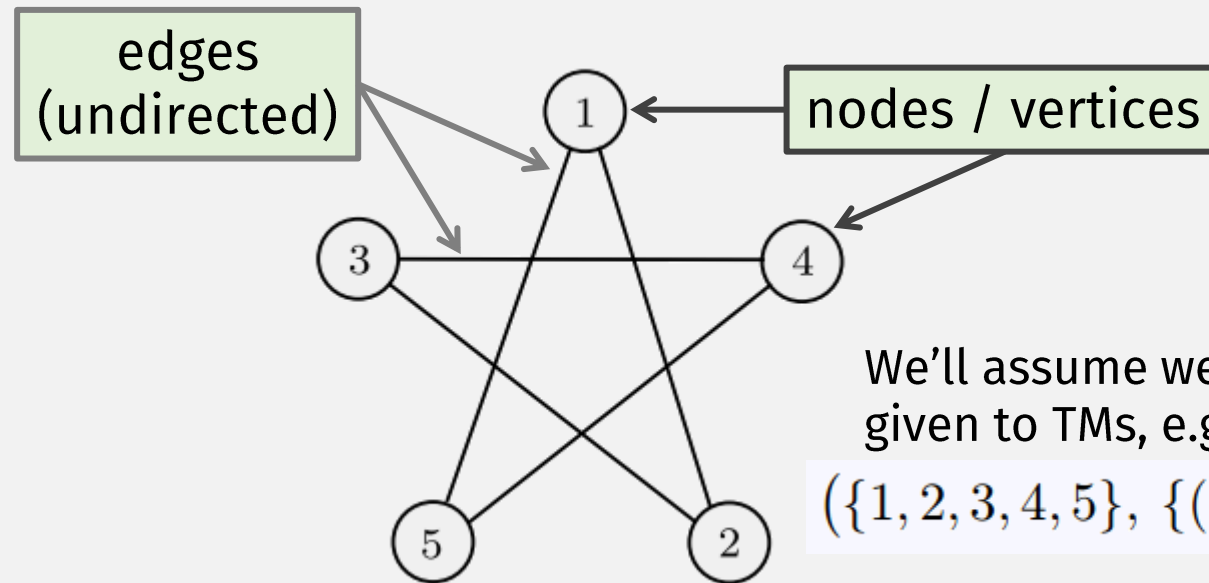
$PATH = \{ \langle G, s, t \rangle \mid G \text{ is a directed graph that has a directed path from } s \text{ to } t \}$

- To prove that a language is in P ...



- ... we must construct a polynomial time algorithm deciding the lang
- A non-polynomial (i.e., exponential, brute force) algorithm:
 - Check all possible paths, and see if any connect s to t

Interlude: Graphs (see Chapter 0)



We'll assume we have some **string encoding**, $\langle G \rangle$, given to TMs, e.g.:

$(\{1, 2, 3, 4, 5\}, \{(1, 2), (2, 3), (3, 4), (4, 5), (5, 1)\})$

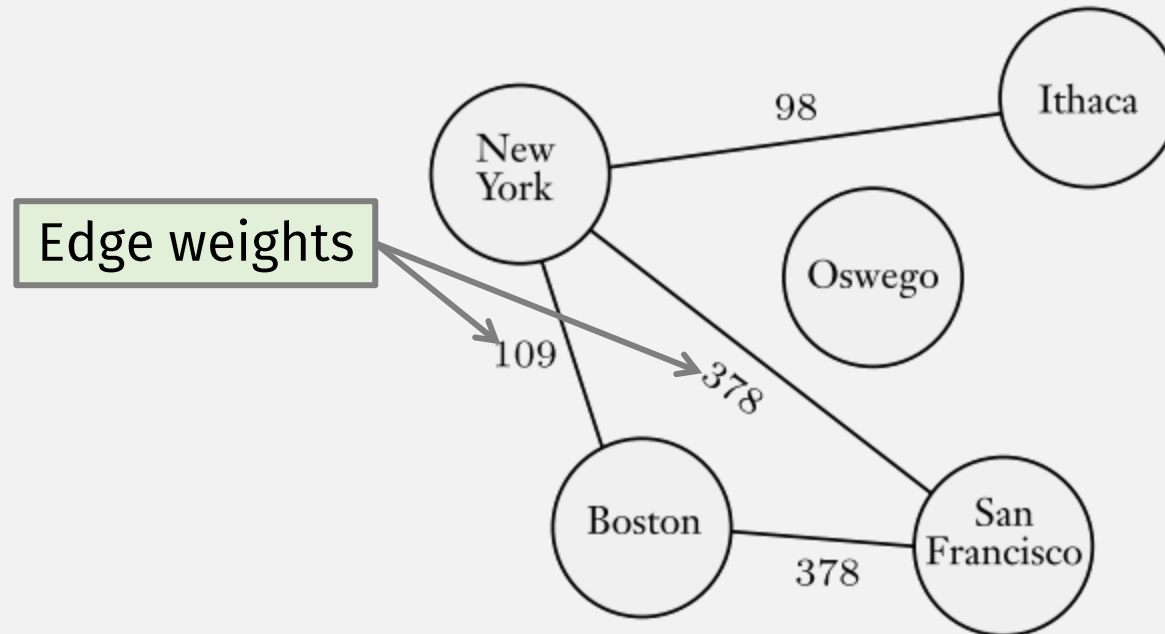
- Edge defined by two nodes (order doesn't matter)
- Formally, a graph = (V, E)
 - V = set of nodes, E = set of edges

Interlude: Graph Encodings

$(\{1, 2, 3, 4, 5\}, \{(1, 2), (2, 3), (3, 4), (4, 5), (5, 1)\})$

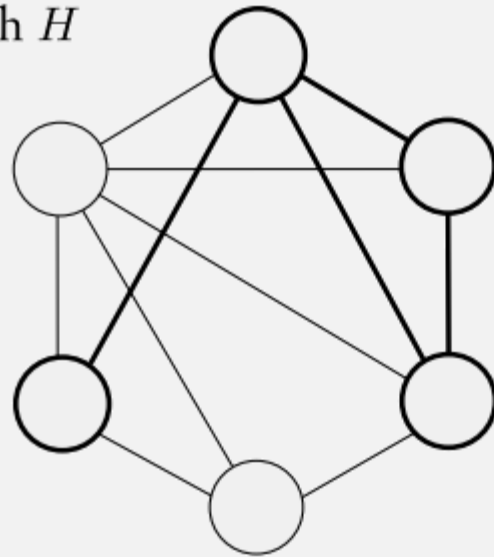
- In graph algorithms, “length of input” n = number of vertices
 - and sometimes number of edges
 - Not number of chars
 - So steps counted in terms of number of vertices
- Given a graph $G = (V, E)$ with $n = |V|$ vertices
- Max edges = $O(|V|^2) = O(n^2)$
- So # vertices + edges is polynomial in length of input
- Algorithm runs in time polynomial in the number of vertices \Leftrightarrow algorithm runs in time polynomial in the length of input

Interlude: Weighted Graphs



Interlude: Subgraphs

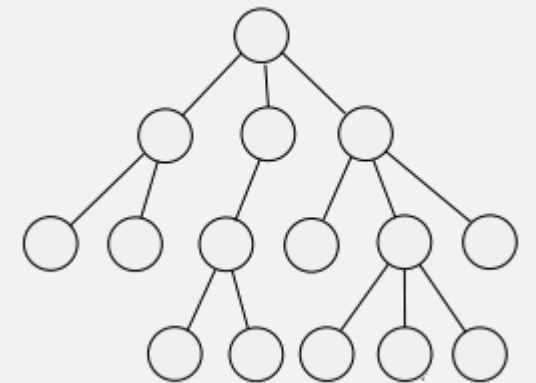
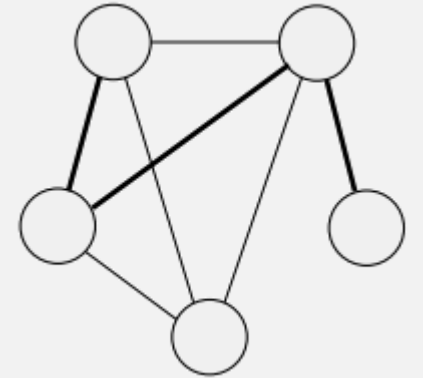
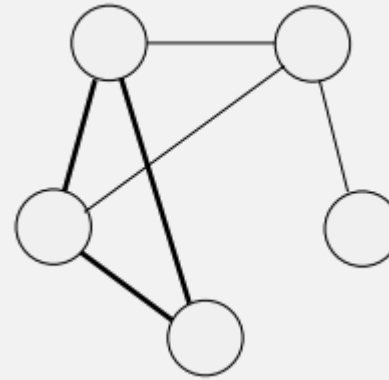
Graph H



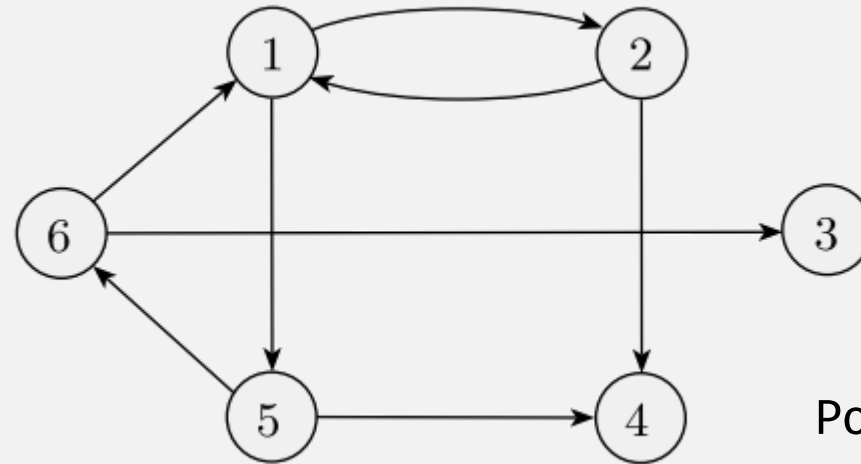
Subgraph G
shown darker

Interlude: Paths and other Graph Things

- Path
 - A sequence of nodes connected by edges
- Cycle
 - A path that starts/ends at the same node
- Connected graph
 - Every two nodes has a path
- Tree
 - A connected graph with no cycles



Interlude: Directed Graphs



Possible **string encoding** given to TMs:

$(\{1,2,3,4,5,6\}, \{(1,2), (1,5), (2,1), (2,4), (5,4), (5,6), (6,1), (6,3)\})$

- Directed graph = (V, E)
 - V = set of nodes, E = set of edges
- An edge is a pair of nodes (u,v) , order now matters
 - u = “from” node, v = “to” node
- A “degree” of a node is the number of edges connected to the node
 - Nodes in a directed graph have both indegree and outdegree

Each pair of nodes
included twice

Check-in Quiz 4/14

On gradescope