## CSE 373: Master method, finishing sorts, intro to graphs

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The tree method: precise analysis
Problem: Need a rigorous way of getting a closed form
We want to answer a few core questions:
How much work does each recursive level do?

1. How many nodes are there on level $?$ ? $(i=0$ is "root" level)
2. At some level $i$, how much work does a single node do? (Ignoring subtrees)
3. How many recursive levels are there?

How much work does the leaf level (base cases) do?

1. How much work does a single leaf node do?
2. How many leaf nodes are there?

## The tree method: precise analysis

How many levels are there, exactly? Is it $\log _{2}(n)$ ?
Let's try an example. Suppose we have $T(4)$. What happens?


Height is $\log _{2}(4)=2$.
For this recursive function, num recursive levels is same as height.
Important: total levels, counting base case, is height +1 .
Important: for other recursive functions, where base case doesn't happen at $n \leq 1$, num recursive levels might be different then height.

## The tree method: precise analysis

We discovered:

```
1. numNodes \((i)=2^{i}\)
2. workPerNode \((n, i)=\frac{n}{2^{1}}\)
3. numLevels \((n) \quad=\log _{2}(n)\)
4. workPerLeafNode \((n)=1\)
5. \(\operatorname{numLeafNodes}(n)=2^{\text {numLevels }(n)}=2^{\log _{2}(n)}=n\)
```

Our formulas:

$$
\begin{aligned}
\text { recursiveWork } & =\sum_{i=0}^{\text {numLevels }(n)} \text { numNodes }(i) \cdot \text { workPerNode }(n, i) \\
\text { baseCaseWork } & =\text { numLeafNodes }(n) \cdot \text { workPerLeafNode }(n) \\
\text { totalWork } & =\text { recursiveWork }+ \text { baseCaseWork }
\end{aligned}
$$

## The tree method: precise analysis

Solve for recursive case:

$$
\begin{aligned}
\text { recursiveWork } & =\sum_{i=0}^{\log _{2}(n)} 2^{i} \cdot \frac{n}{2^{i}} \\
& =\sum_{i=0}^{\log _{2}(n)} n \\
& =n \log _{2}(n)
\end{aligned}
$$

Solve for base case:

$$
\text { baseCaseWork }=\text { numLeafNodes }(n) \cdot \text { workDonePerLeafNode }(n)
$$

$$
=n \cdot 1=n
$$

So exact closed form is $n \log _{2}(n)+n$.

Practice: Let's go back to our old recurrence...

$$
S(n)= \begin{cases}2 & \text { if } n \leq 1 \\ 2 S(n / 3)+n^{2} & \text { otherwise }\end{cases}
$$

## The tree method: practice

```
1. numNodes \((i)=2^{i}\)
2. workPerNode \((n, i)=\frac{n^{2}}{9}\)
3. numLevels \((n) \quad=\log _{3}(n)\)
4. workPerLeafNode( \(n\) ) \(=2\)
5. numLeafNodes \((n) \quad=2^{\text {numLevels }(n)}=2^{\log _{3}(n)}=n^{\log _{s}(2)}\)
```

Combine into a single expression representing the total runtime.

$$
\begin{aligned}
\text { totalWork } & =\left(\sum_{i=0}^{\log _{3}(n)} 2^{i} \cdot \frac{n^{2}}{9^{i}}\right)+2 n^{\log _{3}(2)} \\
& =n^{2} \sum_{i=0}^{\log _{3}(n)} \frac{2^{i}}{9^{\prime}}+2 n^{\log _{3}(2)} \\
& =n^{2} \sum_{j=0}^{\log _{3}(n)}\left(\frac{2}{9}\right)^{i}+2 n^{\log _{3}(2)}
\end{aligned}
$$

## Applying the finite geometric series

With a bunch of effort...

$$
\begin{aligned}
\text { totalWork } & =n^{2} \frac{1-\left(\frac{2}{9}\right)^{\log _{9}(n)+1}}{1-\frac{2}{9}}+2 n^{\log _{3}(2)} \\
& =\frac{9}{7} n^{2}\left(1-\frac{2}{9}\left(\frac{2}{9}\right)^{\log _{9}(n)}\right)+2 n^{\log _{3}(2)} \\
& =\frac{9}{7} n^{2}-\frac{2}{7} n^{2}\left(\frac{2}{9}\right)^{\log _{3}(n)}+2 n^{\log _{3}(2)} \\
& =\frac{9}{7} n^{2}-\frac{2}{7} n^{2} n^{\log _{3}(2 / 9)}+2 n^{\log _{3}(2)} \\
& =\frac{9}{7} n^{2}-\frac{2}{7} n^{2} n^{\log _{3}(2)-2}+2 n^{\log _{3}(2)} \\
& =\frac{9}{7} n^{2}-\frac{2}{7} n^{\log _{3}(2)}+2 n^{\log _{3}(2)} \\
& =\frac{9}{7} n^{2}+\frac{12}{7} n^{\log _{9}(2)}
\end{aligned}
$$

## The tree method: practice



## The finite geometric series

We have: $n^{2} \sum_{i=0}^{\log _{3}(n)}\left(\frac{2}{9}\right)^{i}+2 n^{\log _{3}(2)}$
The finite geometric series identity: $\sum_{i=0}^{n-1} r^{i}=\frac{1-r^{n}}{1-r}$
Plug and chug:

$$
\begin{aligned}
\text { totalWork } & =n^{2} \sum_{i=0}^{\log _{3}(n)}\left(\frac{2}{9}\right)^{j}+2 n^{\log _{3}(2)} \\
& =n^{2} \sum_{i=0}^{\log _{3}(n)+1-1}\left(\frac{2}{9}\right)^{i}+2 n^{\log _{s}(2)} \\
& =n^{2} \frac{1-\left(\frac{2}{9}\right)^{\log _{3}(n)+1}}{1-\frac{2}{9}}+2 n^{\log _{9}(2)}
\end{aligned}
$$

## The master theorem

Is there an easier way?

If we want to find an exact closed form, no. Must use either the unfolding technique or the tree technique.

If we want to find a big- $\Theta$ bound, yes.

## The master theorem

## The master theorem

Suppose we have a recurrence of the following form:

$$
T(n)= \begin{cases}d & \text { if } n=1 \\ a T\left(\frac{n}{b}\right)+n^{c} & \text { otherwise }\end{cases}
$$

Then...

- If $\log _{b}(a)<c$, then $T(n) \in \Theta\left(n^{c}\right)$
- If $\log _{b}(a)=c$, then $T(n) \in \Theta\left(n^{c} \log (n)\right)$
- If $\log _{b}(a)>c$, then $T(n) \in \Theta\left(n^{\log _{b}(a)}\right)$


## The master theorem

Given: Then...
$T(n)= \begin{cases}d & \text { If } \log _{b}(a)<c, \text { then } T(n) \in \Theta\left(n^{c}\right) \\ a T\left(\frac{n}{b}\right)+n^{c} & \text { If } \log _{b}(a)=c, \text { then } T(n) \in \Theta\left(n^{c} \log (n)\right) \\ \log _{b}(a)>c, \text { then } T(n) \in \Theta\left(n^{\log _{b}(a)}\right)\end{cases}$
Sanity check: try checking merge sort.
We have $a=2, b=2$, and $c=1$. We know
$\log _{b}(a)=\log _{2}(2)=1=c$, therefore merge sort is $\Theta(n \log (n))$.
Sanity check: try checking $S(n)=2 S(n / 3)+n^{2}$.
We have $a=2, b=3$, and $c=2$. We know $\log _{3}(2) \leq 1<2=c$, therefore $S(n) \in \Theta\left(n^{2}\right)$.

## The master theorem: intuition

Intuition, the $\log _{b}(a)<c$ case:

1. We do work more rapidly then we divide.
2. So, more of the work happens near the "top", which means that the $n^{c}$ term dominates.

The master theorem: intuition

Intuition, the $\log _{b}(a)>c$ case:

1. We divide more rapidly then we do work.
2. So, most of the work happens near the "bottom", which means the work done in the leaves dominates.
3. Note: Work in leaves is about

$$
d \cdot a^{\text {height }}=d \cdot a^{\log _{b}(n)}=d \cdot n^{\log _{b}(a)} .
$$

Intuition, the $\log _{b}(a)=c$ case:

1. Work is done roughly equally throughout tree.
2. Each level does about the same amount of work, so we approximate by just multiplying work done on first level by the height: $n^{c} \log _{b}(n)$.

## Hybrid sorts

Problem: Quick sort, in the best case, is pretty fast - often a constant factor faster then merge sort. But in the worst case, it's $\mathcal{O}\left(n^{2}\right)$ !

Idea: If things start looking bad, stop using quicksort! Switch to a different sorting algorithm.

## Hybrid sort

A sorting algorithm which combines two or more other sorting algorithms.

## Hybrid sort example: introsort

## Example: Introsort

Core idea: Combine quick sort with heap sort
(Why heap sort? It's also $\mathcal{O}(n \log (n))$, and can be implemented in-place.)

1. Run quicksort, but keep track of how many recursive calls we've made
2. If we pass some threshold (usually, $2\lfloor\lg (n)\rfloor$ )
2.1 Assume we've hit our worst case and switch to heapsort
2.2 Else continue using quick sort

Punchline: worst-case runtime is now $\mathcal{O}(n \log (n))$, not $\mathcal{O}\left(n^{2}\right)$.
Used by various $C_{+}++$implementations, used by Microsoft's .NET framework.

## Adaptive sorts

Observation: Most real-world data is contains "mostly-sorted runs" - chunks of data that are already sorted.

Idea: Modify our strategy based on what the input data actually looks like!
Adaptive sort
A sorting algorithm that adapts to patterns and pre-existing order in the input.

Most adaptive sorts take advantage of how real-world data is often partially sorted.

## Adaptive sorts: Timsort

## Example: Timsort

Core idea: Combine merge sort with insertion sort
(Who's Tim? A Python core developer)

1. Works by looking for "sorted runs".
2. Will use insertion sort to merge two small runs and merge sort to merge two large runs
3. Implementation is very complex - lots of nuances, lots of clever tricks (e.g. detecting runs that are in reverse sorted order, sometimes skipping elements)

Punchline: worst-case runtime is still $\mathcal{O}(n \log (n))$, but best case runtime is $\mathcal{O}(n)$.
Used by Python and Java.

## Counting sort

## Counting sort

- Assumption: Input is a list of ints where every item is between 0 and $k$
- Worst-case runtime: $\mathcal{O}(n+k)$

How would you implement this? Hint: start by creating a temp array of length $k$. Take inspiration from how hash tables work.

## The algorithm:

1. Create a temp array named arr of length $k$.
2. Loop through the input array. For each number num, run $\operatorname{arr}[$ num] $+=1$
3. The temp array will now contain how many times we say each number in the input list.
4. Iterate through it to create the output list.

Other interesting linear sorts:

- Radix sort
- Assumes items in list are some kind of sequence-like thing such as strings or ints (which is a sequence of digits)
- Assumes each "digit" is also sortable
- Sorts all the digits in the 1 s place, then the 2 s place...


## - Bucket sort

- A generalization of counting sort
- Assumes items are randomly and uniformly distributed across a range of possibilities


## What is a graph?

## What is a graph?



In this class, by "graph", we mean the graph data structure.

## Graph: formal definition

## Graph

A graph is a pair $G=(V, E)$, where..

- $V$ is a set of vertices
- $E$ is a set of edges (pairs of vertices)


## Notes:

- Vertices are the circle things, edges are the lines
- The words "node" and "vertex" are synonyms

Graph: formal definition examples

Examples:


$$
\begin{array}{rlr}
V=\{a\} & V=\{b, c, d\} & V=\{e, f, g, h\} \\
E=\{ \} & E=\{(b, c),(c, d)\} & E=\{(e, f),(f, g), \\
& & (g, h),(h, e), \\
& & (e, g),(f, h)\}
\end{array}
$$

## Applications of graphs

## In a nutshell:

Graphs let us model the "relationship" between items.
If that seems like a very general definition, that's because graphs are a very general concept!

## Core insight:

- Graphs are an abstract concept that appear in many different ways
- Many problems can be modeled as a graph problem

Some examples...

Application: Airline flight graph


Questions: What is the cheapest/shortest/etc flight from $A$ to $B$ ? Is the route the airline offering me actually the cheapest route? What happens if a city is snowed in - how can we reroute flights?



Questions: Why does this graph look clustered? Why are all my friends more popular then me? Who do my friends know? If I want to hire somebody to promote my product, who do I pick?
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## Application: Analyzing code



Questions: which files import which ones? Which files are most used and should be optimized? What if two files import each other? If a file has a security vulnerability, how might it propagate?
$\qquad$

## Modeling problems with graph

How would you model the following using graphs? Decide what you think the vertices are and what the edges are:

- Maps (e.g. Google Maps)

Idea: vertices are intersections, edges are roads. How do we model traffic? Paths for cyclists vs cars? One-way roads?

- Web pages

Idea: vertices are webpages, links are edges.

- A running program

Idea: model each statement as vertices, and the next lines it can execute as edges.

- Courses at UW

Idea: model courses as vertices, and pre-requisites as edges.

- A family tree

Idea: model people as vertices, and relations as edges. Or the other way around: model events like birth or divorce as vertices, and make people the edges connecting events?

## Is there a "graph" ADT?

## Question: is there a graph ADT?

Well, what operations belong to the ADT? Hmm, lots of ideas (getEdge(v), reachableFrom(...), centrality (...), etc..)

Observation: It's very unclear what the "standard operations" are
Instead, what we do is think about graphs abstractly, and think about which algorithms are relevant to the problems at hand.

## Directed graphs

## Directed graph

In a directed graph, edges do have a direction: are one-way


Now, $(x, y)$ and $(y, x)$ mean different things.
In-degree of a vertex
The in-degree of $v$ is the number of edges that point to $v$.
Out-degree of a vertex
The out-degree of $v$ is the number of edges that start at $v$.

## Some questions

## Questions:

- In an undirected graph, is it possible to have a vertex with a degree of zero?
Yes.
- In a directed graph, a vertex with an in-degree and out-degree of both zero?
Yes.


## Undirected graphs

Undirected graph
In a undirected graph, edges have no direction: are two-way

$T$ is means that $(x, y) \in E$ implies that $(y, x) \in E$. (Often, we treat these two pairs as equivalent and only include one).

## Degree of a vertex

The degree of some vertex $v$ is the number of edges containing that vertex.

So, the degree is the number of adjacent vertices.

## Self-loops and parallel edges

## Self-loop

A self-loop is an edge that starts and ends at the same vertex.


## Parallel edges

Two edges are parallel if they both start and end at the same vertices.


Whether we allow or disallow self-loops and parallel edges depends on what we're trying to model.

## Simple graph

A graph with no self-loops and no parallel edges.

