No Hw is due in Midtern week.

CSE 421

Dijkstra's Algorithm, Divide and Conquer

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Boiling Water Example

Q: Given an empty bowl, how do you make boiling water?



A: Well, I fill it with water, turn on the stove, leave the bowl on the stove for 20 minutes. I have my boiling water.

Q: Now, suppose you have a bowl of water, how do you make boiling water?

A: First, I pour water away, now I have an empty bowl and I have already solved this!



Lesson: Never solve a problem twice! Related Q: Try to do Max-spanning tree using min-spanning tree









































Disjkstra's Algorithm: Correctness

Prove by induction that throughout the algorithm, for any $u \in S$, the path P_u in the shortest from s to u.

Base Case: This is always true when $S = \{s\}$.

IH: Suppose |S| = k and the claim holds for S

IS: Say v is the k+1-st vertex that we add to S. Let $\{u,v\}$ be last edge on P_{ν} . If P_{ν} is not the shortest path there is a path P to s which is shorter. Consider the first time that P leaves S (with edge $\{x,y\}$). S -> x has weight (at least) d(x)So, $c(P) \ge d(x) + c_{x,v} \ge d(v) = c(P_v)$. A contradiction.



Remarks on Dijkstra's Algorithm

- Algorithm also produces a tree of shortest paths to s following Parent links
- Algorithm works on directed graph (with nonnegative weights)
- The algorithm fails with negative edge weights.
 - e.g., some airline tickets

Why does it fail?



- Dijkstra's algorithm is similar to BFS:
 - Subtitute every edge with $c_e = k$ with a path of length k, then run BFS.



Implementing Dijkstra's Algorithm

Priority Queue: Elements each with an associated key Operations

- Insert
- Find-min
 - Return the element with the smallest key
- Delete-min
 - Return the element with the smallest key and delete it from the data structure
- Decrease-key
 - Decrease the key value of some element

Implementations

Arrays:

- O(n) time find/delete-min,
- O(1) time insert/decrease key

Binary Heaps:

- O(log n) time insert/decrease-key/delete-min, O(1) time find-min

Dijkstra's Algorithm

Runs in $O((n+m)\log n)$.

```
Dijkstra(G, c, s) {
    foreach (v \in V) d[v] \leftarrow \infty //This is the key of node v
   d[s] \leftarrow 0
   foreach (v \in V) insert v onto a priority queue Q
   Initialize set of explored nodes S \leftarrow \{s\}
   while (Q is not empty) { Q(1)
       u \leftarrow delete min element from Q
                                                             O(n) of delete min,
       S \leftarrow S \cup \{u\}
                                                             each in O(log n)
       foreach (edge e = (u, v) incident to u)
            if ((v \notin S) \text{ and } (d[u]+c_e < d[v]))
                d[v] \leftarrow d[u] + c_{e}
                Decrease key of v to d[v].
                Parent(v) \leftarrow
}
                                  O(m) of decrease key,
                                  each runs in O(\log n)
```

Summary (Greedy Algorithms)

- Greedy Stays Ahead: Interval Scheduling, Dijkstra's algorithm
- Structural: Interval Partitioning
- Exchange Arguments: MST, Kruskal's Algorithm,
- Data Structures: Union Find, Heap

Divide and Conquer Approach

Divide and Conquer

Similar to algorithm design by induction, we reduce a problem to several subproblems.

Typically, each sub-problem is at most a constant fraction of the size of the original problem

Recursively solve each subproblem Merge the solutions

Examples:

Mergesort, Binary Search, Strassen's Algorithm,



A Classical Example: Merge Sort



Why Balanced Partitioning?

An alternative "divide & conquer" algorithm:

- Split into n-1 and 1
- Sort each sub problem
- Merge them

Runtime

$$T(n) = T(n-1) + T(1) + n$$

Solution:

$$T(n) = n + T(n - 1) + T(1)$$

= $n + n - 1 + T(n - 2)$
= $n + n - 1 + n - 2 + T(n - 3)$
= $n + n - 1 + n - 2 + \dots + 1 = O(n^2)$

D&C: The Key Idea

Suppose we've already invented Bubble-Sort, and we know it takes n^2

Try just one level of divide & conquer:

Bubble-Sort(first n/2 elements)

Bubble-Sort(last n/2 elements)

Merge results

Time:
$$2T(n/2) + n = n^2/2 + n \ll n^2$$

Almost twice as fast!



D&C approach

- "the more dividing and conquering, the better"
 - Two levels of D&C would be almost 4 times faster, 3 levels almost 8, etc., even though overhead is growing.
 - Best is usually full recursion down to a small constant size (balancing "work" vs "overhead").

In the limit: you've just rediscovered mergesort!

- Even unbalanced partitioning is good, but less good
 - Bubble-sort improved with a 0.1/0.9 split: $(.1n)^2 + (.9n)^2 + n = .82n^2 + n$

The 18% savings compounds significantly if you carry recursion to more levels, actually giving $O(n \log n)$, but with a bigger constant.

• This is why Quicksort with random splitter is good – badly unbalanced splits are rare, and not instantly fatal.

Finding the Root of a Function

Finding the Root of a Function

Given a continuous function f and two points a < b such that $\begin{cases} f(a) \le 0 \\ f(b) \ge 0 \end{cases}$

Find an approximate root of f (a point c where f(c) = 0).

f has a root in [*a*, *b*] by intermediate value theorem

Note that roots of f may be irrational, So, we want to approximate the root with an arbitrary precision!



A Naiive Approch

Suppose we want ϵ approximation to a root.

Divide [a,b] into $n = \frac{b-a}{\epsilon}$ intervals. For each interval check $f(x) \le 0, f(x + \epsilon) \ge 0$ This runs in time $O(n) = O(\frac{b-a}{\epsilon})$ Can we do faster?

b

а

D&C Approach (Based on Binary Search)

```
Bisection(a,b, ε)
    if (b-a) < \epsilon then
       return (a)
    else
       m \leftarrow (a+b)/2
       if f(m) \leq 0 then
          return(Bisection(c, b, \varepsilon))
       else
         return(Bisection(a, c, \epsilon))
```





Time Analysis

Let
$$n = \frac{a-b}{\epsilon}$$

And $c = (a+b)/2$

Always half of the intervals lie to the left and half lie to the right of c

So,

$$T(n) = T\left(\frac{n}{2}\right) + O(1)$$
i.e.,
$$T(n) = O(\log n) = O(\log \frac{a-b}{\epsilon})$$



Correctness Proof

P(k) = "For any *a*, *b* such that $k\epsilon \le |a - b| \le (k + 1)\epsilon$ if f(a)f(b) < 0, then we find an ϵ approx to a root using $\log k$ queries to f"

```
Base Case: P(1): Output a + \epsilon
IH: Assume P(k).
```

IS: Show P(2k). Consider an arbitrary a, b s.t., $2k\epsilon \le |a - b| < (2k + 1)\epsilon$

If $f(a + k\epsilon) = 0$ output $a + k\epsilon$.

If $f(a)f(a + k\epsilon) < 0$, solve for interval $a, a + k\epsilon$ using log(k) queries to f.

Otherwise, we must have $f(b)f(a + k\epsilon) < 0$ since f(a)f(b) < 0and $f(a)f(a + k\epsilon) \ge 0$. Solve for interval $a + k\epsilon$, b.

Overall we use at most log(k)+1=log(2k) queries to f.