

Kruskal's, Dynamic Programming Intro, LCS



1332 Recitation: Week of July 14th

Kruskal's

- A greedy algorithm that finds an MST in a graph
- Only works for undirected graphs (because of the definition of MSTs)
- We will use a Set of edges to store and represent the MST
- Uses a Disjoint-Set data structure
 - Keeps track of a set of elements partitioned into disjoint sets
 - Each element is initially stored in its own set
 - Has two methods
 - Union fuses two sets together so that they become one set
 - Find returns an “id” for the set (if $\text{find}(x) == \text{find}(y)$, x and y are in the same set)
 - We use a Disjoint-Set because it can perform these two methods very fast -> practically $O(1)$

Kruskal's in High-level

1. Add edges to the MST in sorted order by weights (which will be provided to us by a PQ, which initially contains **all** edges)
2. Skip an edge if adding it to the MST would create a cycle
3. Terminate when an MST is found (i.e. when the number of edges in the MST is $2(|V| - 1)$)

Kruskal's - Algorithm

- Initialize a Disjoint-Set of vertices using its constructor (note that each vertex starts in its own set)
- Build a Priority Queue of all edges (BuildHeap!)
- Until MST is found (set has $2(|V| - 1)$ edges) or PQ is empty:
 - Dequeue an edge (u,v)
 - If u and v are not in the same set and thus would not create a cycle (i.e. $\text{find}(u) \neq \text{find}(v)$):
 - Union the sets of u and v (i.e. $\text{union}(u,v)$)
 - Add edge to MST
- Check validity of MST before returning (check if there are $2(|V|-1)$ edges in MST)
- Note that we check whether u and v are in the same set because if they were, adding their edge to the MST would create a cycle
- Example!
- Efficiency: $O(|E|\log|V|)$
 - Comes from all the dequeues and all other parts of the algorithm are dominated by this

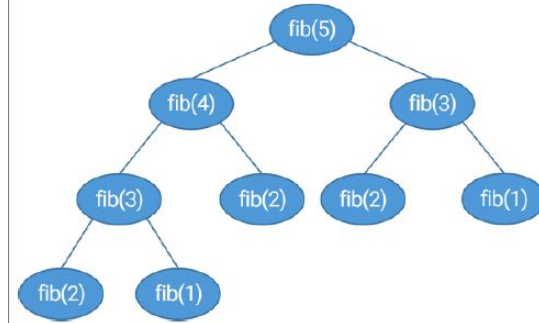
Dynamic Programming Intro

- A strategy for solving a class of problems that can be broken down into smaller repetitive problems that overlap
 - Break the problem into smaller subproblems
 - Solve subproblems and save their results as you solve them
 - If we encounter a subproblem again, just use the saved result
 - Build off the results of subproblems to solve larger instances of the problem
- Typically seen in the context of solving a problem recursively and storing results of recursive calls instead of recomputing
- Using the dynamic programming approach of storing results of sub-problems to avoid re-computation can reduce time complexities from exponential to polynomial time

Fibonacci - Recursive Approach

- The Fibonacci sequence is defined as:
 - $\text{fib}(n) = \text{fib}(n-1) + \text{fib}(n-2)$
 - $\text{fib}(0) = 0, \text{fib}(1) = 1$
 - So $\text{fib}(2) = 1, \text{fib}(3) = 2$
- Recursive approach!

```
public int fib(int n) {  
    if (n == 0) return 0;  
    else if (n == 1) return 1;  
    return fib(n - 1) + fib(n - 2);  
}
```



->does a lot of recomputation, exponential time complexity

->note how fib(3) and fib(2) are called multiple times and would return the same result

Fibonacci - Memoization Approach

- Instead of doing the recursion directly, we first check if we have already computed that case by checking our Map *memo*
 - If we have not yet computed it, we do so and store it in the Map for future
 - We then just return the value stored in the Map
- Note that many nodes in the previous tree will now become leaves and will be computed in constant time

```
Map<Integer, Integer> memo;

public int fib(int n) {
    if (n == 0) return 0;
    else if (n == 1) return 1;
    if (!memo.containsKey(n)) {
        int result = fib(n - 1) + fib(n - 2);
        memo.put(n, result);
    }
    return memo.get(n);
}
```

Fibonacci - Tabulation is EVEN BETTER!

- The approach we'll go over is called "tabulation" -- when we solve a dynamic programming problem by using a table
- We use tabulation if you need to calculate answers to all subproblems, and order of calculation matters
- Tabulation is faster than memoization because it is iterative and requires no overhead space
- The code here can be further optimized to only use a length 2 array (or 2 variables), but this is easier to visualize

```
public int fib(int n) {  
    int[] f = new int[n + 2];  
    f[0] = 0;  
    f[1] = 1;  
    for (int k = 2; k <= n; k++) {  
        f[k] = f[k - 1] + f[k - 2];  
    }  
    return f[n];  
}
```


Longest Common Subsequence (LCS)

- Subsequences :
 - A subset of a string where each character appears in the same order as in the string, but that are not necessarily contiguous
 - Subsequence of BROWN: BRO, BON, BROWN, RWN, O
 - Not subsequences of BROWN: BOR, NWORB, XYZ, NW
- LCS Problem:
 - Given a string X with length n and a string Y of length m, what is the longest subsequence that appears in both strings?

LCS - Algorithm

- Bad! - brute force
 - Iterate over all subsequences in X and check if each subsequence exists in Y
 - Very slow: there are 2^n subsequences in X (each letter can be included or not)
- Good! - dynamic programming
 - Define a 2D array L to store solutions to the subproblems
 - 2D array will be $(m+1) \times (n+1)$ big (we add a row and column for empty strings)
 - Subproblem: $L[i,j]$ will store the length of the LCS of $X[0...i]$ and $Y[0...j]$
 - Set $L[0,k] = 0$ for all $k < m + 1$ and $L[q,0] = 0$ for all $q < n + 1$ -> LCS of any string and an empty string is an empty string
 - The value of each $L[i,j]$ is:
 - If $X[i] == Y[j] \rightarrow L[i-1, j-1] + 1$ (extend the LCS of $X[0...i-1]$ and $Y[0...j-1]$)
 - If $X[i] != Y[j] \rightarrow \max(L[i-1, j], L[i, j-1])$ (use the longest LCS of $X[0...i-1]$ and $Y[0..j]$ or $X[0...i]$ and $Y[0...j-1]$)
 - Both are $O(1)$ operations, fill out $(m*n)$ cells with $O(1)$ operations -> $O(mn)$
- This algorithm only gives us the length of the LCS

LCS - to find the actual LCS

- Set $i = n - 1, j = m - 1$
- Let R be the result string
- If $X[i] == Y[j]$, then $R = R + X[i]; i--; j--;$
- Else if $X[i] != Y[j]$:
 - If $L[i-1, j] \geq L[i, j-1]$, then $i--;$
 - Else $j--$
- Reverse R

LCS - Notes

- Multiple LCSs can be found if you traverse up vs. left when $L[i-1, j] == L[i, j-1]$ while constructing the actual LCS
 - On exams, we will either point out which one to do or tell you to be consistent
- Instead of reversing the string to find the LCS, we could alternatively build the string by adding new chars to front