

DP. Memoization

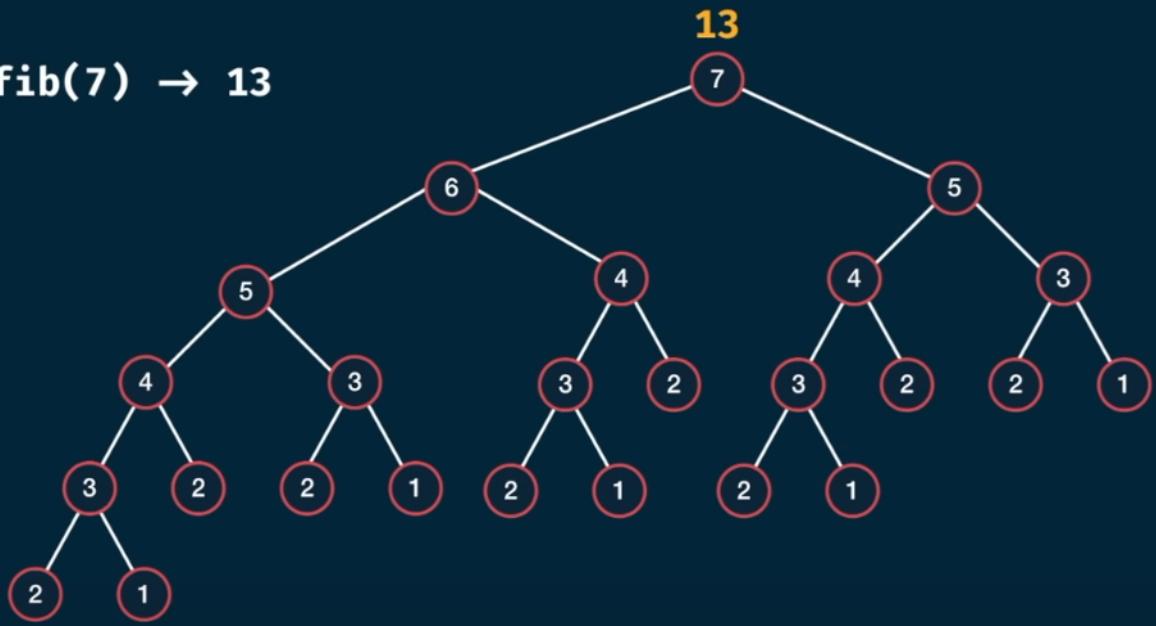
Fibonacci Sequence

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

```
const fib = (n) =>{  
    if(n <= 2) return 1;  
    return fib(n-1) + fib(n-2);  
}  
  
console.log(fib(7));
```

```
1 const fib = (n) => {  
2     if (n <= 2) return 1;  
3     return fib(n - 1) + fib(n - 2);  
4 };
```

fib(7) → 13



- Time complexity of fib function is $O(2^n)$

Time and space complexity of recursive function

- Analysing time complexity of a recursive function
-

```

1  const foo = (n) => {
2    if (n <= 1) return;
3    foo(n - 1);
4  };

```

O(n) time
O(n) space



- Space complexity
 - when we analyse space complexity of recursive functions We should include any of the additional stack space
- for a function

```

void dib(int n){
  if(n <= 1){
    return;
  }
  dib(n-1);
  dib(n-1);
}

```

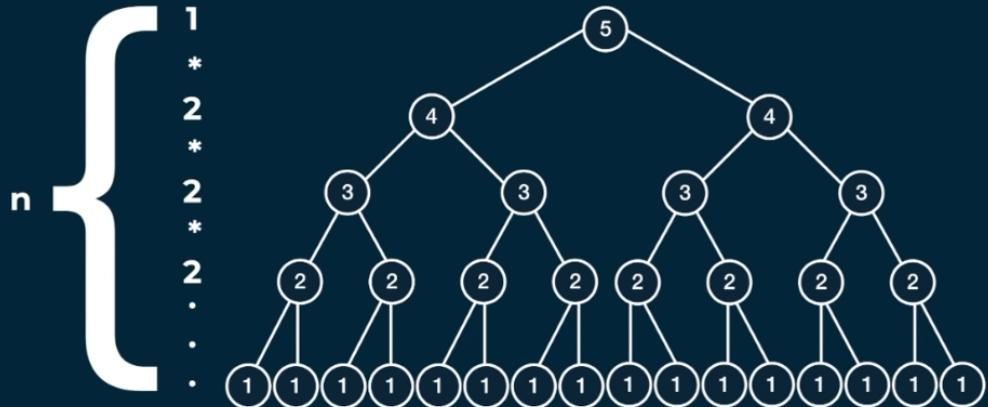
- Height is the distance between the root node and the furthest leaf node
- At every level we are calling the functions twice doubling the number of function calls
- for n levels our time complexity becomes $O(2^n)$

```

1 const dib = (n) => {
2   if (n <= 1) return;
3   dib(n - 1);
4   dib(n - 1);
5 };

```

O(2^n) time



- At any point we use only 5 stacks for `dib(5)`
- Because after reaching five stack calls the functions returns
- it is only after return from left one we enter the right one.

```

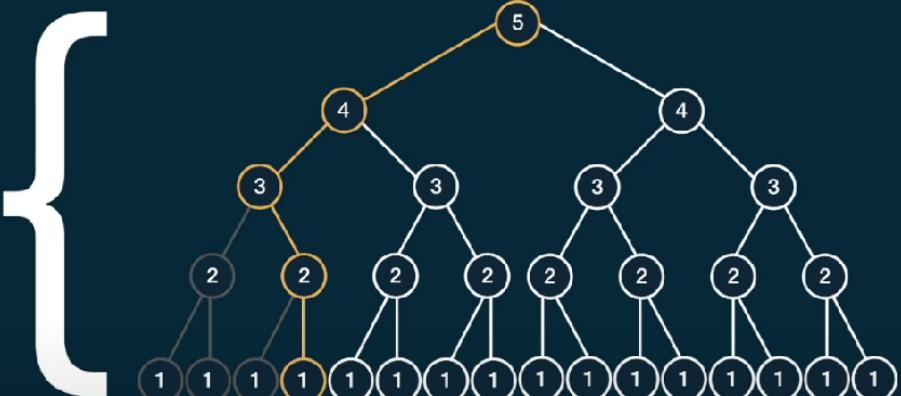
1 const dib = (n) => {
2   if (n <= 1) return;
3   dib(n - 1);
4   dib(n - 1);
5 };

```

O(2^n) time
O(n) space

-

max stack depth of n



- Time complexity is $O(n^2)$ and space complexity is $O(n)$

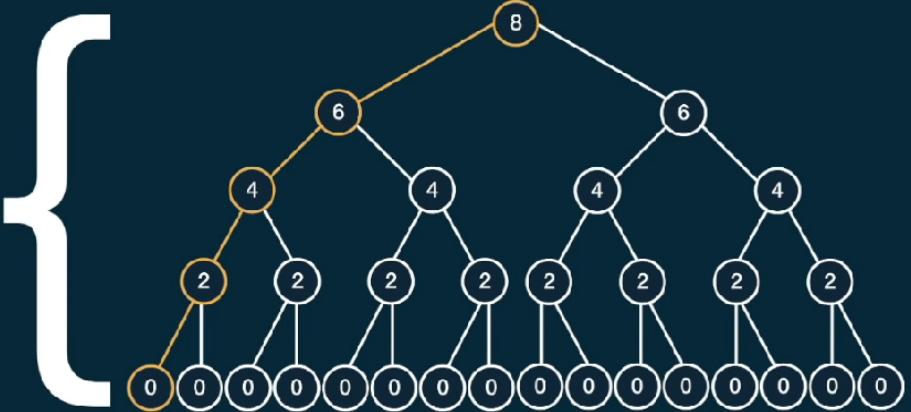
```

1  const lib = (n) => {
2    if (n <= 1) return;
3    lib(n - 2);
4    lib(n - 2);
5  };

```

O(2^n) time
O(n) space

- height of $n/2$



- Our Fibonacci function has $O(2^n)$ time complexity and $O(n)$ space complexity

O(2^n) time
O(n) space

- ```

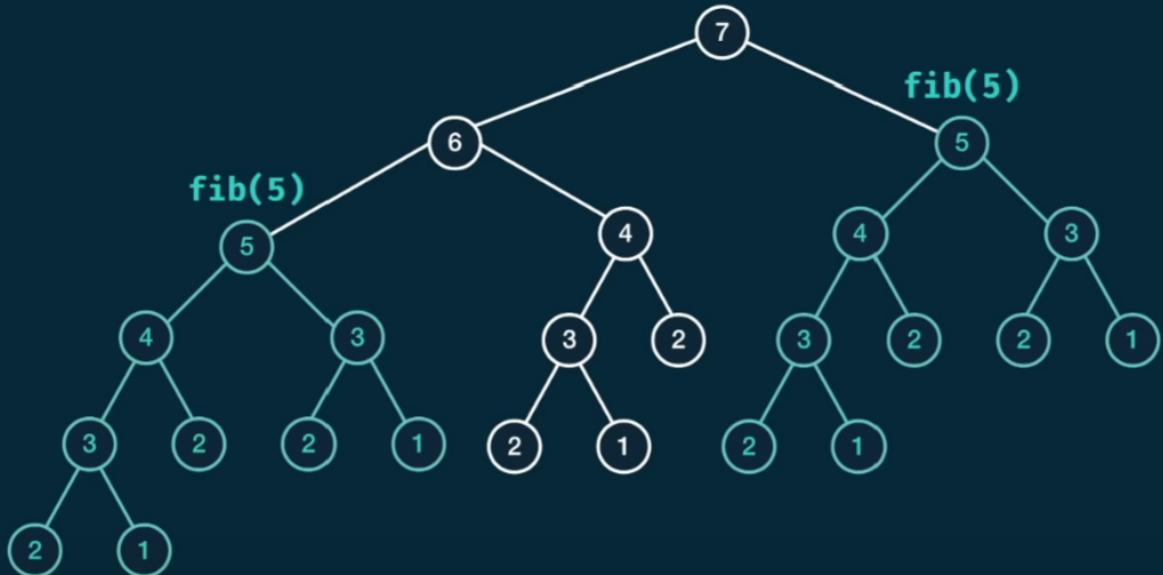
1 const fib = (n) => {
2 if (n <= 2) return 1;
3 return fib(n - 1) + fib(n - 2);
4 };

```

## Dynamic programming definition

# Dynamic Programming

**fib(7)**



- When we have some larger problem, we can decompose into smaller instances of the same problem.  
This concept of break down is called **Dynamic Programming**

## memoization

- We store duplicate subproblems as we can get those results later on.
- To implement a memoization we pick a fast access data structure
- Usually a HashMap or a JavaScript object
- JavaScript passes data to function by reference

### Memoization for Fibonacci

```
int fib(int n){
 // create a memo
 static unordered_map<int, int> memo;
 // find the number
 auto it = memo.find(n);
 if(it != memo.end()){
 return memo[n];
 }

 if(n <= 2){
 return 1;
 }

 // add to the memo
 memo[n] = fib(n-1) + fib(n-2);

 // return from memo
 return memo[n];
}
```

```

const fib = (n, memo = {}) =>{
 // check in memo
 if(n in memo) {
 return memo[n];
 }

 if(n <= 2) {
 return 1;
 }

 memo[n] = fib(n-1, memo) + fib(n-2, memo);
 return memo[n];
}

console.log(fib(30));

```

## Memoized solution recursion tree

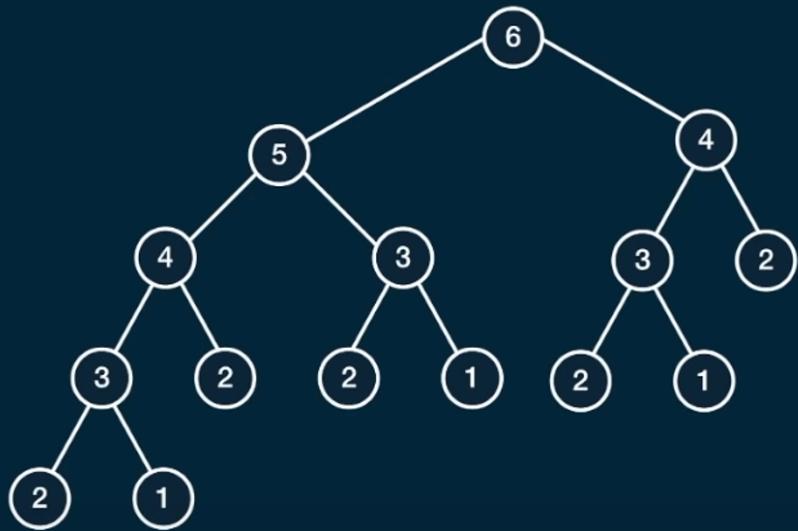
### 1. Without Memorization

```

1 const fib = (n, memo = {}) => {
2 if (n in memo) return memo[n];
3 if (n <= 2) return 1;
4
5 memo[n] = fib(n - 1, memo) + fib(n - 2, memo);
6 return memo[n];
7 };

```

**fib(6)**



### 2. With memoization

```

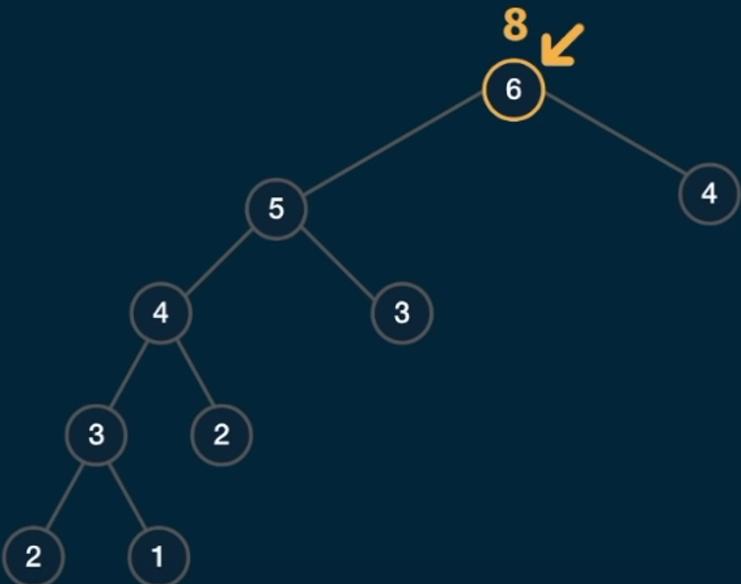
1 const fib = (n, memo = {}) => {
2 if (n in memo) return memo[n];
3 if (n <= 2) return 1;
4
5 memo[n] = fib(n - 1, memo) + fib(n - 2, memo);
6 return memo[n];
7 };

```

**fib(6)**

memo

```
{
 3: 2,
 4: 3,
 5: 5,
 6: 8
}
```



- Overall we have roughly  $2n$  nodes



**fib memoized complexity**  
**O(n) time**  
**O(n) space**

- 

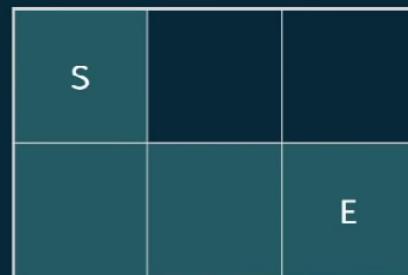
- $\text{fib}(9)$  has this structure
- New time complexity is  $O(n)$  and space complexity is  $O(n)$

# Grid Traveller

- You are a traveller on a 2D Grid. You begin in the top-left corner and your goal is to travel to the bottom right corner. You may only move down or right.
- In how many ways can you travel to the goal on a grid with dimensions m\*n?
- 

`gridTraveler(2,3) → 3`

1. right, right, down
2. right, down, right
3. down, right, right



- Problems like this.
  - Always start with the smallest possible scenario
    - `gridTraveler(1,1) -> 1` since start is already at end
      - do nothing
    - `gridTraveler(0,1) -> 0`
      - invalid
    - `gridTraveler(3,3)`
      -
  - These can be thought of as base cases and can be used to construct the larger solution
- for a `gridTraveler(3,3)`
  - As we move down the problem reduces to `gridTraveler(2,3)`
- 

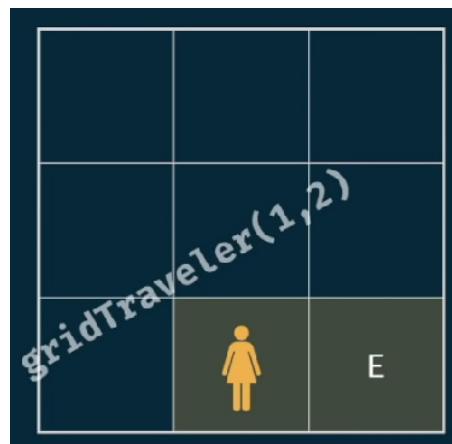
`gridTraveler(3,3) → ?`



- **gridTraveler(3,3) → ?**



- 

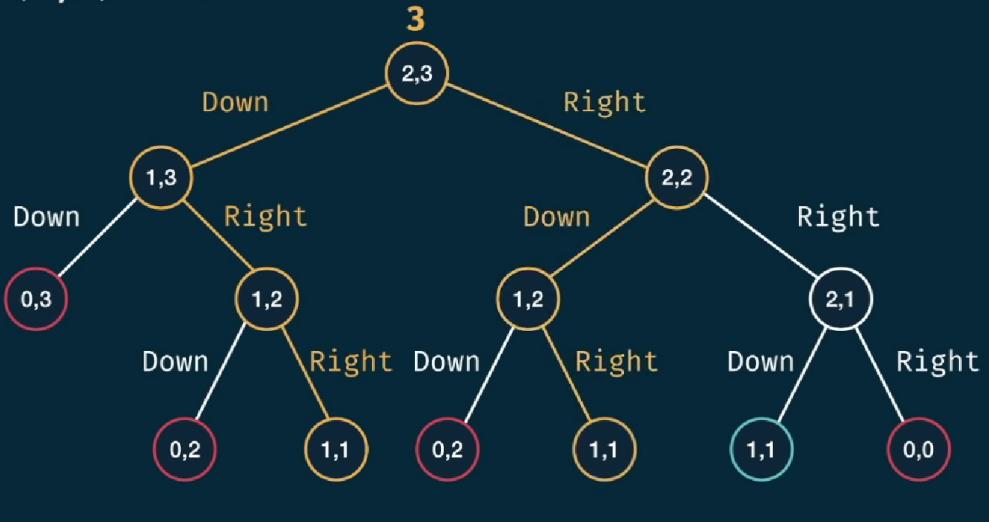


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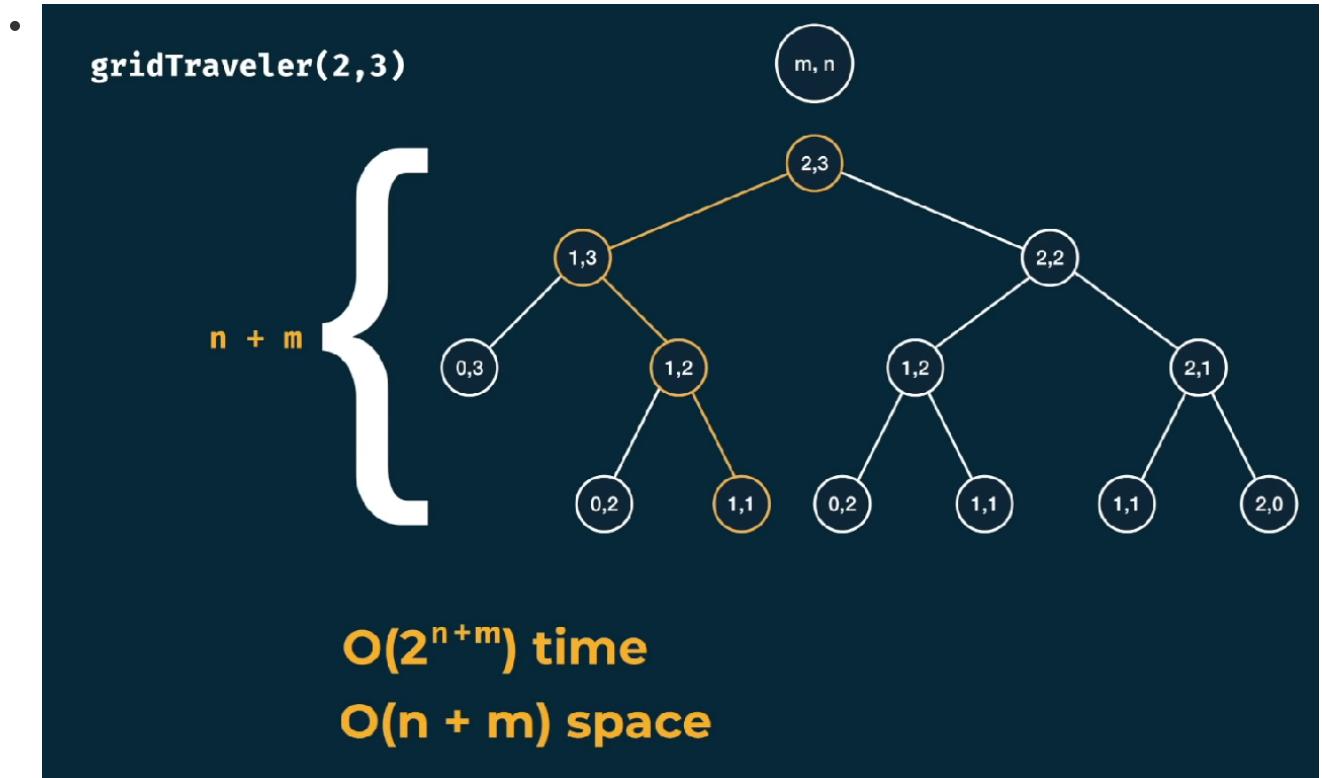


- Any problem that looks like recursion always visualize a tree
- At every recursive call we are either reducing by 1 row or by 1 column
- From the tree structure we can already tell which tree nodes were formed by which combination of moves.
- which combination gives a positive answer and which moves lead us to dead end
-

`gridTraveler(2,3) → 3`



- The tree looks like a binary tree because at every point there are two possible ways to traverse the the grid either down or right
- This tree is now composed of two different factors :  $m$  and  $n$ .
- We must take in account the other parameter
- The height of the tree
  - To do that we travel furthest and see the base case
  - There are two base cases
    - either one of them is 0 and we return
    - Or both are 1 and we count as 1
  - There are  $m+n$  levels



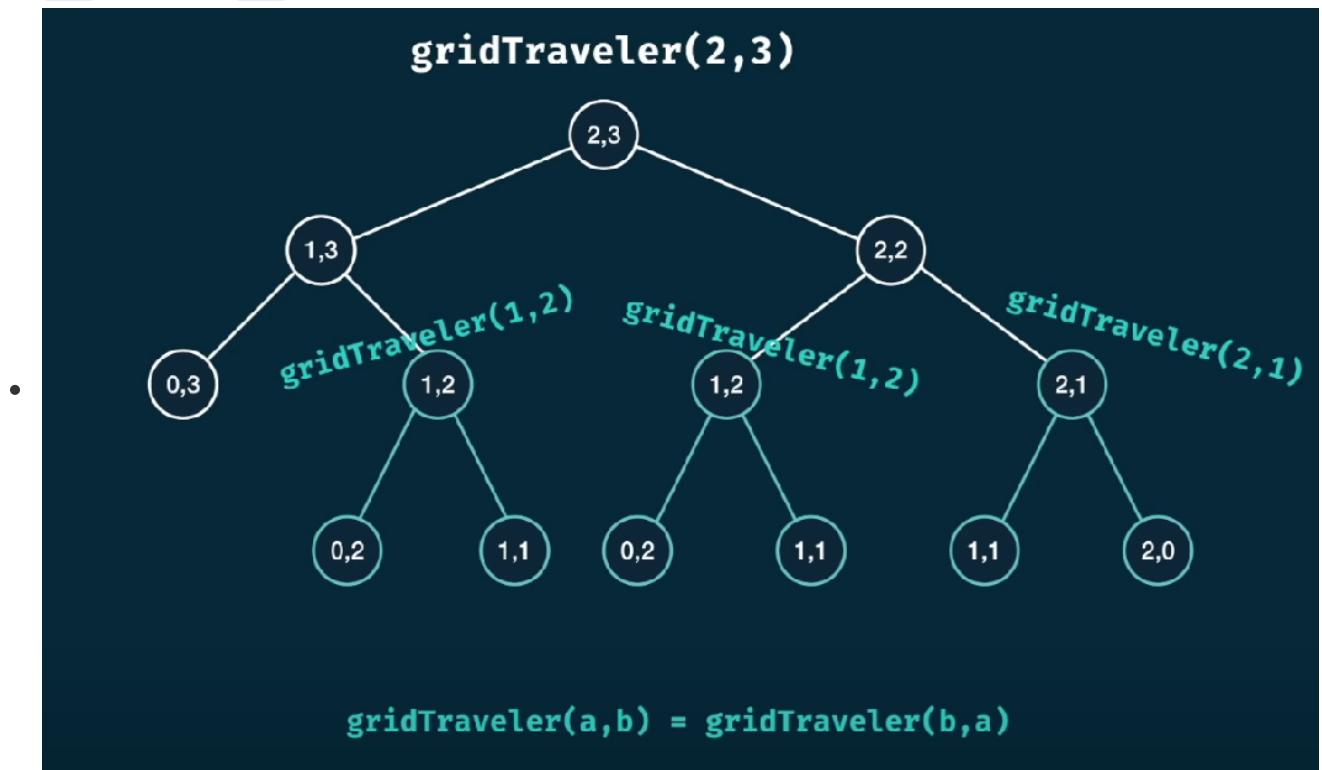
## Brute force recursive solution

```
const gridTraveller = (m,n) =>{
 if(m === 1 && n === 1) return 1;
 if(m === 0 || n === 0) return 0;

 return gridTraveller(m-1, n) + gridTraveller(m, n-1);
}
```

## Making improvements to counting number of ways we can reduce the recursive

- We can notice there are duplicate subtrees
- $[2,1]$  is same as  $[1,2]$



## Strategy

- To solve these recursive dynamic programming problems we need a brute force solution to the original recursive problem
- A well formed recursion

## For memoization

- We take the exact return recursive statement and assign it to the `memo[key]`
- and return that value based key

## Solution

```
const gridTraveller = (m, n, memo={}) =>{
 let key = m + ',' + n;
 if(key in memo) return memo[key];
 if(m === 1 && n === 1) return 1;
 if(m === 0 || n === 0) return 0;

 memo[key] = gridTraveller(m-1, n, memo) + gridTraveller(m, n-1, memo);
 return memo[key];
}
```

- A pair cannot be used as key in unordered\_list

```
// wrong solution
int gridTraveller(int n, int m){
 // create memo table
 static unordered_map<pair<int, int>, int> memo;
 // find the pair
 auto it = memo.find({n, m});
 if(it != memo.end()){
 return memo[{n,m}];
 }

 if(m == 1 && n == 1) return 1;

 if(m == 0 || n == 1) return 0;

 memo[{n,m}] = gridTraveller(m-1, n) + gridTraveller(m, n-1);
}
```

```
int gridTraveller(int n, int m){
 // crate a key
 string key = to_string(n) + " , "+ to_string(m);
 // create memo table
 static unordered_map<string, int> memo;
 // find the pair
 auto it = memo.find(key);
 if(it != memo.end()){
 return memo[key];
 }

 if(m == 1 && n == 1) return 1;

 if(m == 0 || n == 0) return 0;

 memo[key] = gridTraveller(m-1, n) + gridTraveller(m, n-1);
 return memo[key];
}
```

}

## Space and Time complexity

- For a case of `gridTraveler(4,3)`
  - `m: {0,1,2,3,4}`
  - `n: {0,1,2,3}`
- Total number of nodes possible `m*n`
- Memoized version has time and space complexity of
  - Time : `O(m*n)`
  - Space: `O(n+m)`

## Memoization Recipe

### Two Step process

#### 1. Make it work

1. Visualize the problem as a tree
2. Implement the tree using recursion (brute force)
3. Test it (should give correct results, it may be slow)

#### 2. Make it efficient

1. Add a memo object (key represent argument to the function, values represent return values), it must be shared among all recursive calls
2. Add a base case to return memo values
3. Store return values into the memo

**Don't try to implement an efficient algorithm from get go. Get a brute force working solution using recursion then implement using memoization.**

- first look for correctness in your solution it could be brute force or anything. As long as it can get you correct results
- Then once that is achieved then try to optimize it.
- Leaves of the tree are the base cases

## canSum

- Write a function `canSum(targetSum, numbers)` that takes in a targetSum and an array of numbers as an arguments.
- The function should return a boolean indicating whether or not it is possible to generate the targetsum using numbers from the array
- You can use an element of the array as many times as needed
- You may assume that all input numbers are nonnegative

```
bool canSum(int target, vector<int> & v){
 cout << "canSum(" << target << ")" << endl;
 if(target == 0){
```

```

 return true;
 }

 if(target < 0){
 return false;
 }

 for(auto a: v){
 int remainder = target -a;
 if(canSum(remainder, v)){
 return true;
 }
 }

 return false;
}

```

```

const canSum(targetSum, numbers){
 if(targetSum === 0) return true;
 if(targetSum < 0) return false;

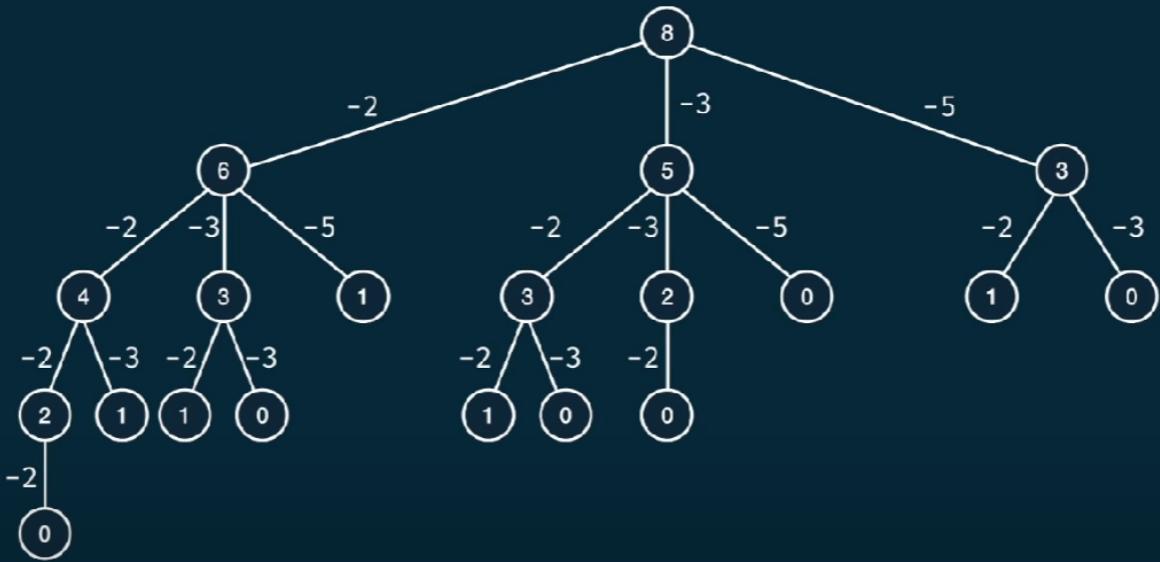
 for(let num of numbers){
 const remainder = targetSum - num;
 if(canSum(remainder, numbers) === true){
 return true;
 }
 }

 return false;
}

```

## Time and space complexity of brute force

`canSum(8, [2, 3, 5]) → true`



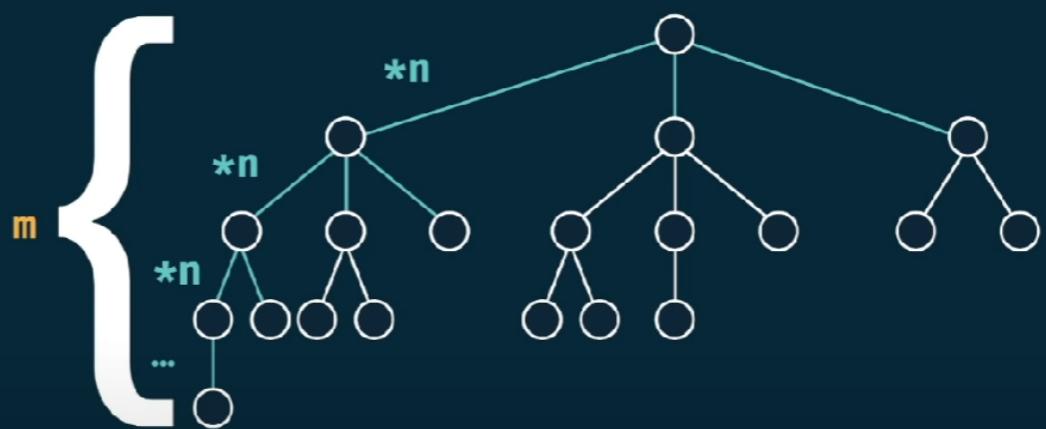
- Two factors determine the size of the tree
  - $m$  = target sum
  - $n$  = array length
- Height of tree
  - Even if we subtract 1 at every step to reach the target sum the maximum height would be  $\text{targetsum}$
  - $m$  is the height of the tree
- Branches
  - The maximum branching factor is the number of elements in the array:  $n$

$m$  = target sum

$n$  = array length

**$O(n^m)$  time**

**$O(m)$  space**



- The time complexity becomes  $O(n^m)$
- Space used will be maximum height of the tree  $O(m)$

## Memoized solution

```
bool canSum(int target, vector<int> & v){

 static unordered_map<int, bool> memo;

 unordered_map<int, bool>::iterator it = memo.find(target);

 if(it == memo.end()){
 return memo[target];
 }

 if(target == 0){
 return true;
 }

 if(target < 0){
 return false;
 }

 for(auto a: v){
 int remainder = target - a;
 memo[remainder] = canSum(remainder, v);
 if(memo[remainder]){
 return true;
 }
 }

 memo[target] = false;
 return false;
}
```

```
const canSum = (target, numbers) =>{
 if(target == 0) return true;
 if(target < 0) return false;

 for(let num of numbers){
 const remainder = target - num;
 if(canSum(remainder, numbers) === true){
 return true;
 }
 }

 return false;
}
```

```
const canSum = (target, numbers, memo = {}) =>{
```

```

if(target in memo) return memo[target];

if(target == 0) return true;
if(target < 0) return false;

for(let num of numbers){
 const remainder = target - num;
 memo[remainder] = canSum(remainder, numbers, memo);

 if(memo[remainder]){
 return true;
 }
}

memo[target] = false;
return memo[target];
}

```

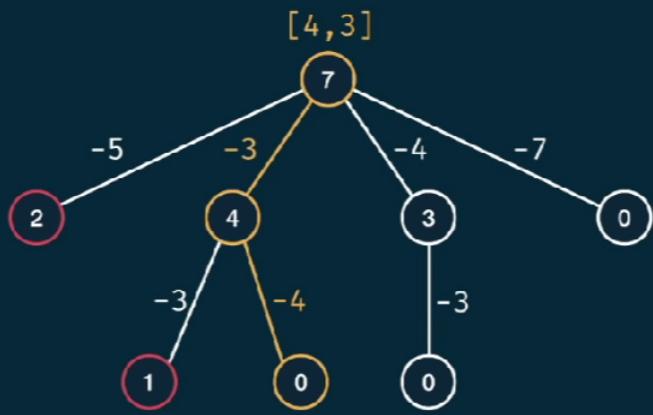
## Time and space complexity of memoized solution

- $m$  the target sum is the height of the array
- $n$  the size of the array determines the branches
- Time complexity is  $O(n*m)$ 
  - We still need to branch  $n$  times to collect the values for the memo object
- Space complexity is height  $O(m)$

## howSum(targetSum, numbers)

- Write a function `howSum(targetSum, numbers)` that takes in a targetsum and an array of numbers as arguments
- The function should return an array containing any combination of elements that add up to exactly the targetSum.
- If there is no combination that adds up to the targetSum, then return null.
- We can return as soon as we find a combination
- Example
  - a
    - `howSum(8, [2,3,5]) -> [2,2,2,2]`
    - `howSum(8, [2,3,5]) -> [3,5]`
  - b
    - `howSum(7, [2,4]) -> null`
  - c

- `howSum(0, [1,2,3])-> []`  
`howSum(7, [5, 3, 4, 7])`



## Recursive solution

```
const howSum = (target, numbers) =>{

 if(target === 0){
 return [];
 }

 if(target < 0){
 return null;
 }

 for(let c of numbers){
 const remainder = target - c;
 const result = howSum(remainder, numbers);
 if(result !== null){
 return [...result, c];
 }
 }

 return null;
}

console.log(howSum(7, [2,3]));
```

```
// c++ 17 and onwards
// g++ -std=c++17 index.cpp -o index
#include <iostream>
#include <unordered_map>
#include <algorithm>
#include <vector>
#include <utility>
#include <string>
```

```

#include <optional>

using namespace std;

optional<vector<int>> howSum(int sum, vector<int> &v){
 if(sum < 0){
 return nullopt;
 }

 if(sum == 0){
 return vector<int>();
 }

 for(auto a: v){
 int remainder = sum -a;
 optional<vector<int>> result = howSum(remainder, v);
 if(result.has_value()){
 result->push_back(a);
 return *result;
 }
 }

 return nullopt;
}

int main(){
 vector<int> v = {2, 3};

 optional<vector<int>> v2 = howSum(7, v);
 if(v2.has_value()){
 for(auto a: *v2){
 cout << a;
 }
 }
 return 0;
}

```

- You can use `optional<vector<int>>` to either return a `vector<int>` or return a `nullopt` which represents a null value.
- `optional<vector<int>> v2` can be checked for if there is a value by using `.has_value()`

## space and time complexity for howSum

- `m` = target sum
- `n` = length of vector
- `time` =  $O(n^m * m)$

- Target sum determines the height of the tree
- vector length determine number of branches at every height
- Space complexity
  - $m$  is the number of max stack size at any point
  - space =  $O(m)$

## memoized solution

```
const howSum = (target, numbers, memo = []) =>{

 if(target in memo){
 return memo[target];
 }

 if(target === 0){
 return [];
 }

 if(target < 0){
 return null;
 }

 for(let c of numbers){
 const remainder = target - c;
 const result = howSum(remainder, numbers, memo);
 if(result !== null){
 memo[target] = [...result, c];
 return memo[target];
 }
 }

 memo[target] = null;
 return null;
}
```

```
optional<vector<int>> howSum(int sum, vector<int> &v){

 static unordered_map<int, optional<vector<int>>> memo;

 auto it = memo.find(sum);

 if(it != memo.end()){
 return it->second;
 }

 if(sum < 0){
 return nullopt;
 }
```

```

 if(sum == 0){
 return vector<int>();
 }

 for(auto a: v){
 int remainder = sum -a;
 optional<vector<int>> result = howSum(remainder, v);
 if(result.has_value()){
 result->push_back(a);
 memo[sum] = *result;
 return memo[sum];
 }
 }

 memo[sum] = nullopt;
 return nullopt;
}

int main(){
 vector<int> v = {15, 2};

 optional<vector<int>> v2 = howSum(3000, v);
 if(v2.has_value()){
 for(auto a: *v2){
 cout << a << " ";
 }
 }
 return 0;
}

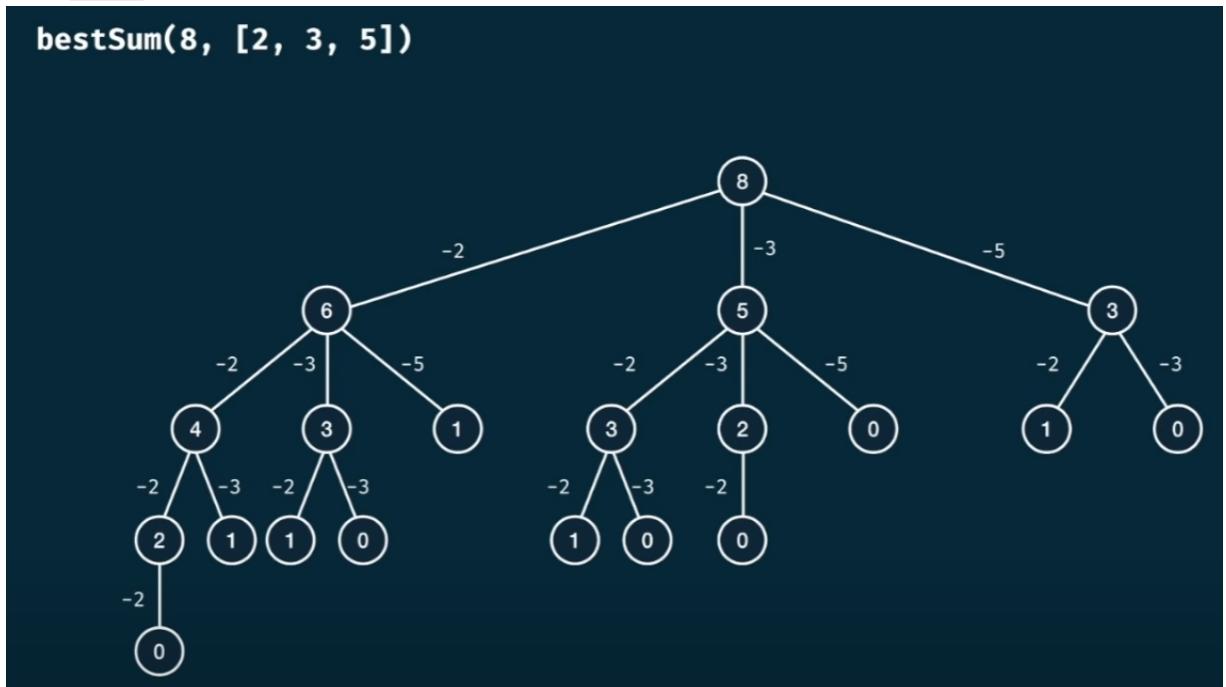
```

## memoized version time and space

- Time
  - Tree height is  $n$
  - $m$  is the vector size
  - copying the array with cost  $m$
  - So total time is  $O(n*m*m)$
  - Time:  $O(n*m^2)$
- Space
  - We must consider maximum number of recursive call at any point:  $m$
  - And all unique values of the memo. The maximum length is  $m$
  - Space  $O(m^2)$

## bestSum(targetSum, numbers)

- Write a function `bestSum(targetSum, numbers)` that takes in a targetSum and an array of numbers as arguments.
- The function should return an array containing the shortest combination of numbers that add up to exactly the targetSum
- If there is a tie for the shortest combination, you may return any one of the shortest.
- Eg
  - `bestSum(7, [5,3,4,7])`
    - `[3,4]`
    - `[7]` result
  - `bestSum(8, [2,3,5])`
    - `[2,2,2,2]`
    - `[2,3,2]`
    - `[3,5]`
  - **`bestSum(8, [2, 3, 5])`**



- **Recursive solution**

```

const bestSum = (targetSum, numbers) =>{
 if(targetSum === 0){
 return [];
 }

 if(targetSum < 0){
 return null;
 }

 let lastResult = null;

 for(let num of numbers){
 let remainder = targetSum - num;
 }
}

```

```

 let result = bestSum(remainder, numbers);
 if(result !== null){
 let current = [...result, num];
 if(lastResult === null || current.length < lastResult.length){
 lastResult = current;
 }
 }
 }

 return lastResult;
}

```

```

optional<vector<int>> bestSum(int targetSum, vector<int>& numbers){
 if(targetSum == 0){
 return vector<int>();
 }

 if(targetSum < 0){
 return nullopt;
 }

 optional<vector<int>> minimumLength = {};// nullopt

 for(auto a: numbers){

 int remainder = targetSum -a;

 optional<vector<int>> result = bestSum(remainder, numbers);

 if(result.has_value()){
 result->push_back(a);
 if(minimumLength == nullopt || result->size() < minimumLength->size()){
 // * minimumLength = * result; does not work
 minimumLength = result; // only this works
 }
 }
 }

 //printOpt(minimumLength);
 return minimumLength;
}

```

## Time and space complexity

- Space
  - height of tree is  $m$  targetsum
  - for every recursive call we are making an array of size  $m$
  - Space :  $O(m)$

- Time complexity
  - Height of tree is  $m$
  - Array size:  $n$
  - For every call we are iteration through an array of size  $m$  because it is the size of the array we are returning so if any number is 1 then we could have an array of just 1  $m$  times
  - $O(n^m * m)$

## Memoized solution

```
const bestSum = (targetSum, numbers, memo = {}) =>{
 if(targetSum in memo){
 return memo[targetSum];
 }

 if(targetSum === 0){
 return [];
 }

 if(targetSum < 0){
 return null;
 }

 let lastResult = null;

 for(let num of numbers){
 let remainder = targetSum - num;

 let result = bestSum(remainder, numbers, memo);
 if(result !== null){
 let current = [...result, num];
 if(lastResult === null || current.length < lastResult.length){
 lastResult = current;
 }
 }
 }

 memo[targetSum] = lastResult;
 return lastResult;
}

console.log(bestSum(100, [1,2,5,25]));
console.log(bestSum(7, [5, 3,4,7]));
```

```
optional<vector<int>> bestSum(int targetSum, vector<int>& numbers){

 static unordered_map<int, optional<vector<int>>> memo;

 unordered_map<int, optional<vector<int>>>::iterator it = memo.find(targetSum);
```

```

 if(it != memo.end()){
 return memo[targetSum];
 }

 if(targetSum == 0){
 return vector<int>();
 }

 if(targetSum < 0){
 return nullopt;
 }

 optional<vector<int>> minimumLength = {};// nullopt

 for(auto a: numbers){

 int remainder = targetSum -a;

 optional<vector<int>> result = bestSum(remainder, numbers);

 if(result.has_value()){

 result->push_back(a);

 if(minimumLength == nullopt || result->size() < minimumLength->size()){

 // * minimumLength = * result; does not work
 minimumLength = result; // only this works
 }
 }
 }

 memo[targetSum] = minimumLength;
 return minimumLength;
}

```

## Time and space complexity

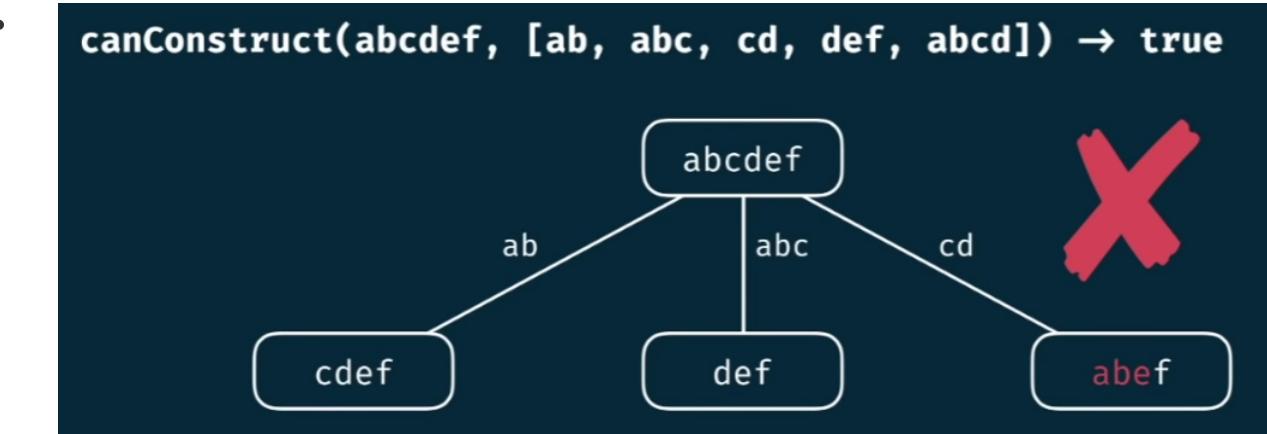
- Time complexity  $O(m*n)$
- Complexity:  $O(m^2 * n)$

## canSum vs howSum vs bestSum

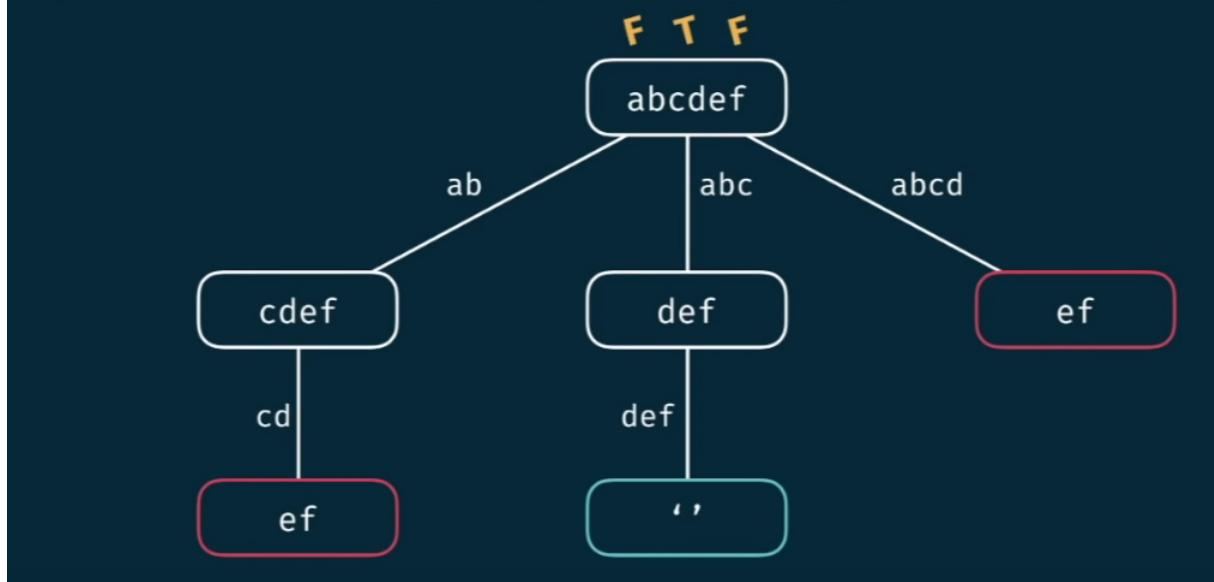
- canSum -> Decision Problem
- howSum -> Combinatoric Problem
- bestSum -> Optimization Problem

## canConstruct

- Write a function `canConstruct(target, wordBank)` that accepts a target string and an array of strings
- The function should return a boolean indicating whether or not the `target` can be constructed by concatenating elements of the `wordBank` array
- You may reuse element of `wordBank` as many times as needed
- `canConstruct(abcdef, [ab, abc, cd, def, abcd]) -> true`
  - `abc+def`
- base case
  - `canConstruct('', [cat, dog, mouse]) -> true`
- Do Not take from the middle only from front or back

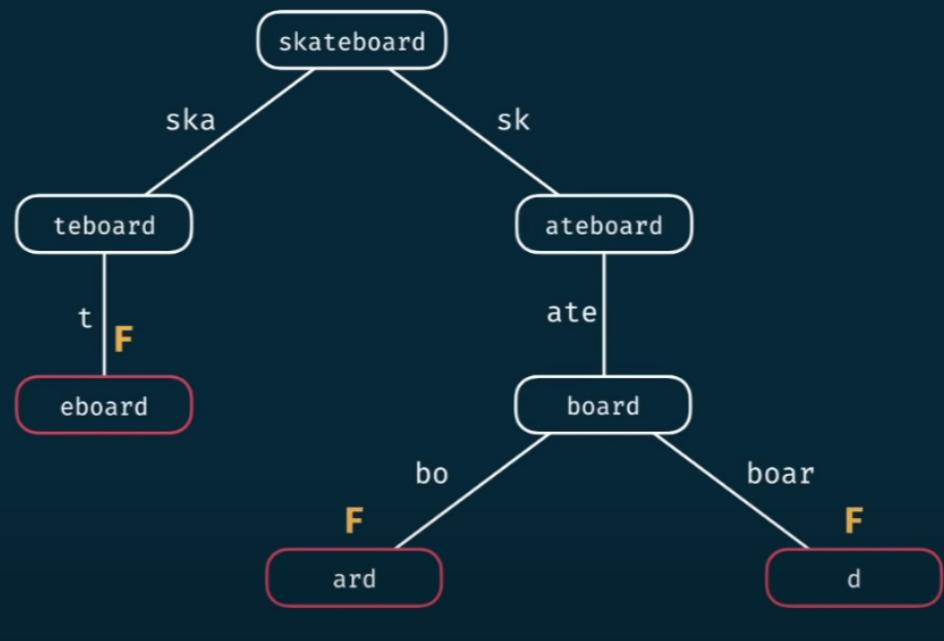


- **`canConstruct(abcdef, [ab, abc, cd, def, abcd]) → true`**



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**skateboard**  
[ bo, rd, ate, t, ska, sk, boar ] → false



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