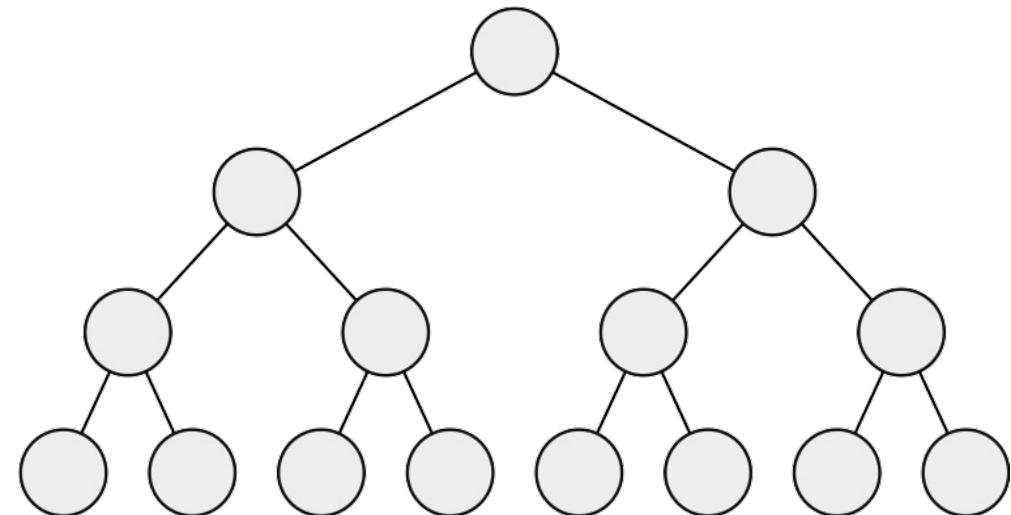


# Competitive Programming

Memoization

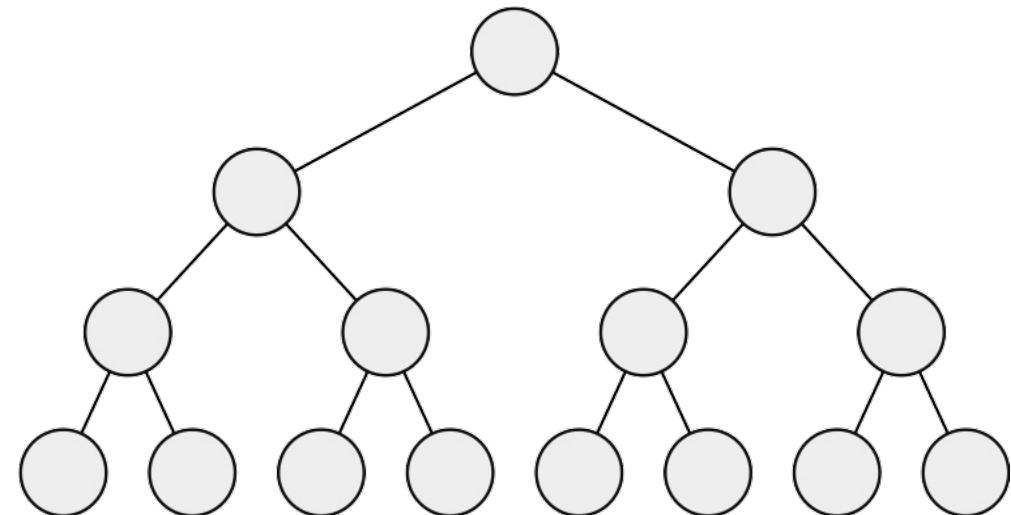
# Duplicate Subproblems

- Often, we can solve a problem by breaking it down into smaller subproblems.
  - Dynamic programming is like this, where the subproblems are smaller, simplified problem instances.
  - Search is like this, where the subproblems involve trying different partial solutions.
- Solving a problem can look like traversing or exploring a tree.



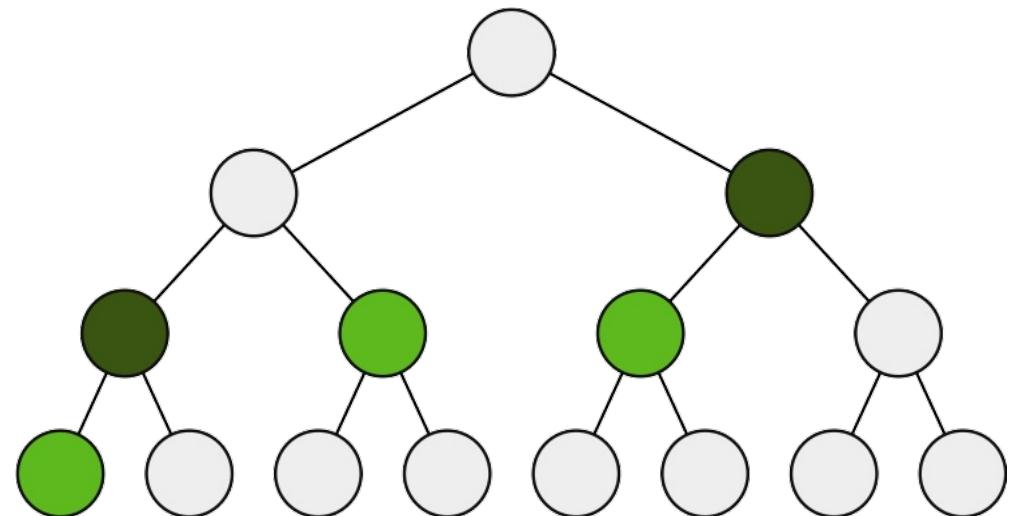
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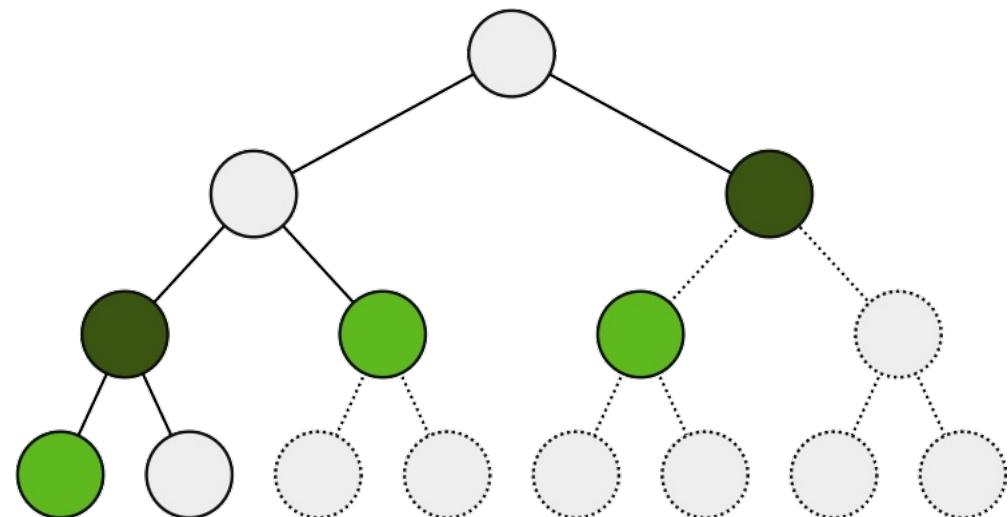
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# Duplicate Subproblems

- The same subproblem can show up along multiple branches.
- Saving a partial solution could avoid duplicate work.
- In some cases, it could turn an exponential-time solution into polynomial.



# Fibonacci Numbers Recursively

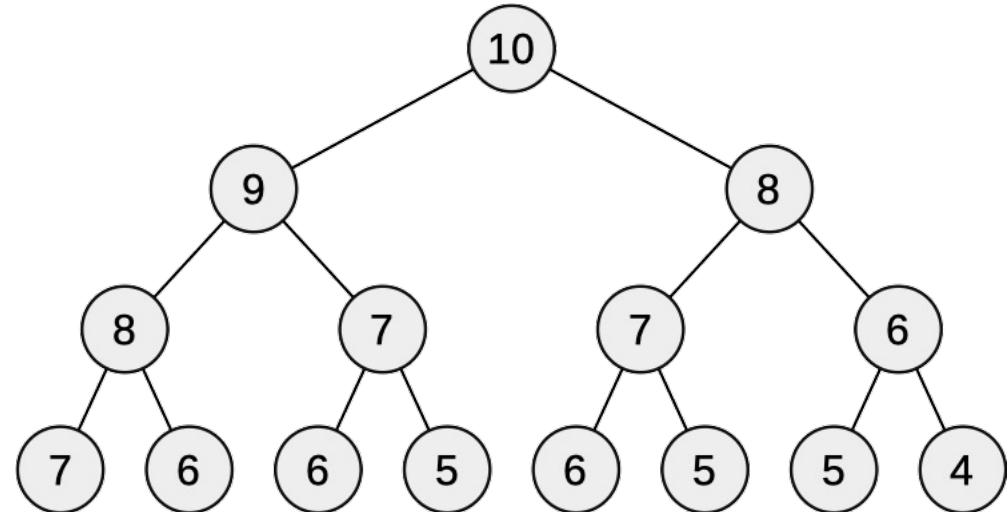
- It's easy to define Fibonacci numbers recursively.

```
int fib( n )
    if ( n < 2 )
        return 1
    return fib( n-1 ) + fib( n-2 )
```

To compute a  
Fibonacci number ..

Compute the previous  
two Fibonacci numbers.

- But this is a bad way to implement it.



# Recursive with a Cache

- A cache of previous solutions would speed this up a lot.

```
map< int, int > cache;

int fib( n )
    if ( n < 2 )
        return 1
    if ( cache.containsKey( n ) )
        return cache[ n ];
    int val = fib( n-1 ) + fib( n-2 );
    cache[ n ] = val;
    return val;
}
```

Let me check  
my cache.

Save this solution for the  
future.

