Problem Description

made up of several units, and at each unit, there may be a stone present. The frog can only land on stones and is not allowed to land in the water.

In this problem, we are tasked with determining whether a frog is able to cross a river by hopping from stone to stone. The river is

The stones are depicted as a list of integer positions, in sorted ascending order, representing their place in the river. The challenge is to figure out if the frog can make it to the last stone, starting from the first stone. The frog's first jump is always 1 unit. If the frog's last jump covered k units, the next jump can either be k - 1, k, or k + 1 units. It's important to note that the frog can only move forward, not backward.

stone without falling into the water at any point. Intuition

The main objective is to navigate through the given list of stones and find a sequence of jumps that will lead the frog to the final

To solve the problem, we have to account for multiple possibilities of jump sequences that the frog can take, which suggests a need for exploring various jump lengths dynamically as the frog makes its way from the first to the last stone. This kind of problem is well-

DFS Approach: With the DFS solution, we explore all the possible jump lengths from each stone. We start from the first stone and recursively try every possible next jump length (k-1, k, k+1). If at any point we can jump to a stone that allows us to eventually reach the last stone, we return true. We employ memoization to store the results of subproblems (i.e., if a certain jump from a particular

suited for depth-first search (DFS) or dynamic programming (DP) approaches.

the current position i with the last jump of k units.

stone is possible or not), which prevents us from recalculating the same scenarios multiple times, thus optimizing efficiency. Dynamic Programming Approach: The DP solution sets up a table where each entry f[i][k] represents whether it's possible to reach stone i with a last jump of k units. We initialize the first stone's entry as true because the frog starts there. Then we systematically fill out the table: for each stone and each potential jump length, we check if we could have gotten there from an

earlier stone with a valid jump length. The tabulated results are reused to infer about the subsequent stones. Both approaches fundamentally examine which stones are reachable from any given stone and under what conditions, ultimately producing a result indicating whether the last stone can be reached or not. **Solution Approach**

The solution we are looking at employs two different strategies: Hash Table with Memoization and Dynamic Programming (DP). Each has its own set of data structures and patterns.

This approach utilizes a recursive Depth-First Search (DFS) method enhanced by a memorization technique to efficiently solve the problem.

Hash Table + Memoization:

• Algorithm: • We use a hash table, denoted as pos, to map each stone's value to its index, which allows us to quickly check if a stone

exists at a certain position. • We implement a recursive function dfs(i, k) that returns true if the frog is capable of reaching the end of the stones from

Data Structures:

Dynamic Programming:

- has successfully reached the end. ○ Otherwise, we explore all feasible jumps from the current stone - the frog can jump k-1, k, or k+1 units forward.
 - We check each of these jumps to see if a stone exists at the new position if so, we make a recursive call to dfs with the new position index and the jump distance. • The answer from subsequent dfs calls is stored using memoization to avoid re-computation, thereby optimizing the solution.

∘ In the function dfs(i, k), if the current index i is equal to the last index n-1 of stones, we return true indicating that the frog

- A hash table (dictionary in Python) is used to quickly determine if a stone exists at a particular jump distance. A memoization cache, which the @cache decorator in Python provides, is used to store the outcomes of the recursive dfs calls.
- Patterns: Memoization pattern is applied here to store already computed results of certain subproblems, avoiding redundant
- computation and reducing time complexity.

The Dynamic Programming approach constructs a DP table where each cell f[i][k] indicates the possibility of reaching the i-th

stone using a jump of k units.

Algorithm:

starts at the first stone with a jump of 0 units. Iterate through all stones and possible jump lengths to populate this table. For each stone i and each jump length k, we check if there is a way to reach i from a previous stone with the previous jump

lengths k-1, k, or k+1. If such a way exists, f[i][k] is set to true.

Remember, the frog can start at the first stone and the first jump is always 1 unit.

jump from the previous stone can be made with a range of k-1, k, or k+1.

using dynamic programming to avoid recalculations.

This function is cached to store the results of subproblems.

for next_jump in range(jump_size - 1, jump_size + 2):

The length of the stones list is stored for easy access.

If the current index is the last stone, then the frog can cross

if next_jump > 0 and index + next_jump in stone_positions:

If none of the possible jumps lead to a crossing, return False.

35 # print(can_cross) # Outputs: True or False based on whether the frog can cross or not

// Determines if it is possible to cross the river by jumping on the given stones

if dfs(stone_positions[index + next_jump], next_jump):

the stone at index 3 with a jump of 2, and the process continues in that manner.

Initialize a 2D table f for the DP, where most of the elements are false except f[0][0] which should be true, as the frog

Finally, check whether any jump length at the last stone entry in the DP table is true, which indicates that the river can be crossed.

Data Structures:

the solution incrementally.

Patterns: Dynamic Programming pattern is employed to break down the problem into smaller overlapping subproblems and build up

A 2D array or list of lists in Python is used to keep track of the dynamic programming state.

The time complexity for both approaches is $0(n^2)$, and the space complexity is also $0(n^2)$, where n is the number of stones. For larger values of n, these approaches might be computationally intense due to the square nature of their complexity.

Using the Hash Table + Memoization approach: 1. We create a hash table as described in the algorithm to map stone positions to their index: pos = {0:0, 1:1, 3:2, 5:3, 6:4,

2. We then implement the dfs(i, k) function, starting at the first stone with index i = 0 and initial jump length k = 0 (this is

Let's consider an example where the stones are placed at the following positions in the river: [0, 1, 3, 5, 6, 8, 12, 17].

3. Checking at i = 0, dfs(0, 0) is called, and we look for stones that can be reached with a jump of 1 (k + 1 since k = 0 initially).

position index 3.

8:5, 12:6, 17:7}.

Example Walkthrough

4. Our next stone is at position 1, and there's indeed a stone there, so we call dfs(1, 1).

5. At i = 1, with k = 1, we have possible jump lengths of 0 (k-1), 1 (k), and 2 (k+1). Only a jump of 2 leads to the next stone at

because the first jump will always be 1, so the last jump length k that brought us to the start is 0).

prevent recalculating scenarios for each stone position and jump length combination. 7. The process continues until we reach the last stone or determine that the frog cannot reach the end. In this case, by following the jumps from position indices 0 -> 1 -> 3 -> 4 -> 5 -> 6 -> 7, the frog can successfully make it across to the last stone.

6. We repeat the process, calling dfs(3, 2). The recursion continues to explore jumps, storing results in the memoization cache to

2. We iterate over each stone position i and consider all possible k values that could reach this stone, checking whether a valid

3. As we fill the table, we find that the frog can reach the stone at index 1 with a jump of 1. Moving on, we find the frog can reach

1. We set up a 2D DP table f with dimensions $n \times n$ initialized with False values, except for f[0][0] which is set to True.

any f[7][k] is True for any k, the frog can cross the river. In our example, the frog can indeed cross the river as there will be True entries indicating valid jumps to the last stone.

In this small example, both approaches demonstrate the frog's ability to cross the river by carefully examining each possible jump

and either using a hash table with memoization to store results of recursive DFS calls or using a DP table to iteratively track the

5. At the end, we check f[n-1][k] for any k to see if the frog can reach the last stone with k jumps. For the last stone at index 7, if

4. We keep building the table based on previous jumps that can reach the current stone until we've considered all stones.

Import the required lru_cache decorator from functools to cache expensive recursive calls from functools import lru_cache class Solution: def canCross(self, stones):

If a valid stone is found, recurse from the position of that stone with the size of the next jump.

A helper function is defined for Depth-First Search (DFS) to determine if the frog can cross

Loop through the potential next jump sizes: jump_size-1, jump_size, jump_size+1

Check if the next jump is possible (jump size must be greater than 0)

and the stone to jump on exists in the stone positions dictionary.

27 # A dictionary (hash map) to store the position of each stone for O(1) look-up. 28 stone_positions = {stone: idx for idx, stone in enumerate(stones)} 29 # Kickstart the recursive DFS with the first stone and initial jump size 0. 30 return dfs(0, 0) 31 32 # Example usage:

```
stone reachable states with diverse jump lengths.
Python Solution
```

@lru_cache(maxsize=None)

return False

def dfs(index, jump_size):

return True

if index == last_stone_index:

return True

last_stone_index = len(stones) - 1

34 # can_cross = sol.canCross([0,1,3,5,6,8,12,17])

9

10

11

12

13

14

15

16

17

18

19

20

21

22

23

24

25

26

36

33 # sol = Solution()

Java Solution

1 class Solution {

Using the Dynamic Programming approach:

```
private Boolean[][] memo; // memoization array to store results of subproblems
 3
        private Map<Integer, Integer> positionMap = new HashMap<>(); // maps stone's value to its index
        private int[] stones; // array of stones
 4
        private int numOfStones; // number of stones
 6
        // Determines if you can cross the river by jumping on stones
        public boolean canCross(int[] stones) {
 8
           numOfStones = stones.length;
 9
10
           memo = new Boolean[numOfStones];
11
           this.stones = stones;
           // populate the position map
12
           for (int i = 0; i < numOfStones; ++i) {</pre>
13
14
                positionMap.put(stones[i], i);
15
16
           // start DFS from stone 0 with jump size 0
17
           return dfs(0, 0);
18
19
20
       // Uses DFS to see if a crossing is possible from the current stone with a given jump size
21
        private boolean dfs(int index, int lastJumpSize) {
22
           // if we are at the last stone, crossing is successful
23
           if (index == num0fStones - 1) {
24
               return true;
25
26
           // if we have visited this state, return the stored result
           if (memo[index][lastJumpSize] != null) {
27
28
               return memo[index][lastJumpSize];
29
30
           // try to jump to the next stone with jump sizes: k-1, k, or k+1
           for (int currentJumpSize = lastJumpSize - 1; currentJumpSize <= lastJumpSize + 1; ++currentJumpSize) {</pre>
31
32
               if (currentJumpSize > 0) { // cannot jump back or stay
33
                    int nextPosition = stones[index] + currentJumpSize; // next stone's position
34
                    // check if there is a stone at the next position
35
                   if (positionMap.containsKey(nextPosition) && dfs(positionMap.get(nextPosition), currentJumpSize)) {
36
                       // if we can cross starting from the next stone, mark the current state as true
                        return memo[index][lastJumpSize] = true;
37
38
39
40
41
           // if none of the jumps work, mark the current state as false
42
           return memo[index][lastJumpSize] = false;
43
44 }
```

for (int i = 0; i < n; ++i) { 15 positionIndexMap[stones[i]] = i; 16 17 18 19 // Depth-first search function that uses memoization

C++ Solution

#include <vector>

2 #include <unordered_map>

bool canCross(vector<int>& stones) {

vector<vector<int>> memo(n, vector<int>(n, -1));

unordered_map<int, int> positionIndexMap;

// Maps stone positions to their indices for quick lookup

int n = stones.size();

#include <functional>

#include <cstring>

6 class Solution {

public:

45

10

11

12

13

14

```
function<bool(int, int)> dfs = [&](int currentIndex, int lastJump) -> bool {
 20
                 // Base case: if the current index is the last one in stones array, the frog can cross
 21
                 if (currentIndex == n - 1) {
 22
 23
                     return true;
 24
                 // If the current state has already been computed, return the stored value
 25
                 if (memo[currentIndex][lastJump] != -1) {
 26
                     return memo[currentIndex][lastJump];
 27
 28
 29
                 // Attempt to jump to the next stone with the last jump minus one, the same, or plus one
 30
                 for (int nextJump = lastJump - 1; nextJump <= lastJump + 1; ++nextJump) {</pre>
                     if (nextJump > 0 && positionIndexMap.count(stones[currentIndex] + nextJump)) {
 31
 32
                         if (dfs(positionIndexMap[stones[currentIndex] + nextJump], nextJump)) {
                             // If a valid jump is found, store the result and return true
 33
 34
                             return memo[currentIndex][lastJump] = true;
 35
 36
 37
 38
                 // If no valid jump can be found, store the result and return false
                 return memo[currentIndex][lastJump] = false;
 39
 40
             };
 41
 42
             // Start DFS from the first stone with an initial jump size of 0
 43
             return dfs(0, 0);
 44
 45
    };
 46
Typescript Solution
     function canCross(stones: number[]): boolean {
         const totalStones = stones.length;
         const positionToIndex: Map<number, number> = new Map();
         // Map each stone's position to its index for quick access
  5
         for (let i = 0; i < totalStones; ++i) {</pre>
  6
             positionToIndex.set(stones[i], i);
  8
  9
         // Memoization table, -1 indicates state has not been computed yet
 10
 11
         const memo: number[][] = new Array(totalStones).fill(0).map(() => new Array(totalStones).fill(-1));
 12
 13
         // Helper DFS function to determine if the frog can reach the end
 14
         const dfs = (index: number, lastJump: number): boolean => {
 15
             // Reached the last stone
             if (index === totalStones - 1) {
 16
 17
                 return true;
 18
```

// f is the memoization table where f[i][j] represents the possibility to reach stone i with the last jump of j units

20 21 22

```
// Return cached result if this state has been computed
            if (memo[index][lastJump] !== -1) {
                return memo[index][lastJump] === 1;
 23
 24
 25
            // Iterate through all possible next jumps (k - 1, k, k + 1)
 26
            for (let jumpDistance = lastJump - 1; jumpDistance <= lastJump + 1; ++jumpDistance) {</pre>
                if (jumpDistance > 0 && positionToIndex.has(stones[index] + jumpDistance)) {
 27
                    const nextIndex = positionToIndex.get(stones[index] + jumpDistance)!;
 28
 29
                    if (dfs(nextIndex, jumpDistance)) {
 30
                        memo[index][lastJump] = 1;
 31
                        return true;
 32
 33
 34
 35
 36
            // Cache the result (0 for false)
 37
            memo[index][lastJump] = 0;
 38
            return false;
        };
 39
 40
 41
        // Start DFS from the first stone with 0 as the last jump distance
 42
        return dfs(0, 0);
 43 }
 44
Time and Space Complexity
```

The space complexity is also 0(n^2) because of the caching. For each unique state (i, k), the result of the computation is stored to prevent re-computation. Considering that there can be up to n unique values for i and n unique values for k, the space required for caching these results can take up to n * n space in the worst-case scenario.

The time complexity of the given code is $0(n^2)$. This is because, in the worst case, the dfs function is called for every stone, and at

each stone, it attempts three different jump distances (k - 1, k, k + 1). The @cache decorator ensures that each combination of i

values (since the furthest a frog can jump from stone i is to stone i + k), which results in a total of n * n unique states.

(stone index) and k (last jump distance) is calculated only once. There are at most n different i values and potentially n different k