
Each element in the array represents your maximum jump length at that position. Determine if you are able to reach the last index.

Example 1:

Example 2:

Output: true

Input: nums = [2,3,1,1,4]

Input: nums = [3,2,1,0,4]Output: false Explanation: You will always arrive at index 3 no matter what. Its maximum jump length

Explanation: Jump 1 step from index 0 to 1, then 3 steps to the last index.

Constraints:

• 1 <= nums.length <= 3 * 10^4 • 0 <= nums[i][j] <= 10^5

Solution

problem is a 4 step process:

Naming

- We call a position in the array a "good index" if starting at that position, we can reach the last index.
- "good index".

1. Start with the recursive backtracking solution

4. Apply final tricks to reduce the time / memory complexity

requirements.

6

7

8 9

15 16 17

18

19

Approach 1: Backtracking This is the inefficient solution where we try every single jump pattern that takes us from the first position to

- until last index is reached. When stuck, backtrack.
- Java 1 public class Solution {
- public boolean canJumpFromPosition(int position, int[] nums) { if (position == nums.length - 1) { 4 return true; 5

10 11 12 13 14 return false;

Intuitively, this means we always try to make the biggest jump such that we reach the end as soon as possible The change required is: **С**ору Java 1 // Old for (int nextPosition = position + 1; nextPosition <= furthestJump; nextPosition++)</pre> for (int nextPosition = furthestJump; nextPosition > position; nextPosition--) For instance, in the example below, if we start from index **0**, jump as far as possible and reach **1**, jump as far as possible and reach **6**. By doing so, we determine that **0** is a GOOD index in 3 steps. 0 2 3 5 6 Index 1 4 5 2 2 0 0 nums To illustrate the worst case, where this optimization has no effect, take the example below. Index 6 cannot be reached from any position, but all combinations will be tried.

Approach 2: Dynamic Programming Top-down Top-down Dynamic Programming can be thought of as optimized backtracking. It relies on the observation that once we determine that a certain index is good / bad, this result will never change. This means that we can store the result and not need to recompute it every time. Therefore, for each position in the array, we remember whether the index is good or bad. Let's call this array memo and let its values be either one of: GOOD, BAD, UNKNOWN. This technique is called memoization².

An example of a memoization table for input array nums = [2, 4, 2, 1, 0, 2, 0] can be seen in the

indices 2, 3 or 4 and eventually reach last index (6), but we can do that from indices 0, 1, 5 and (trivially) 6.

diagram below. We write G for a GOOD position and B for a BAD one. We can see that we cannot start from

2

2

В

3

В

4

0

В

5

2

G

6

0

G

Сору

1. If it is known then return True / False 2. Otherwise perform the backtracking steps as before 3. Once we determine the value of the current index, we store it in the memo table

int furthestJump = Math.min(position + nums[position], nums.length - 1); 13 for (int nextPosition = position + 1; nextPosition <= furthestJump; nextPosition++) {</pre> 14 if (canJumpFromPosition(nextPosition, nums)) { 15 memo[position] = Index.GOOD; 16 17 return true; 18 } 19 20 memo[position] = Index.BAD; 21

Approach 3: Dynamic Programming Bottom-up Top-down to bottom-up conversion is done by eliminating recursion. In practice, this achieves better performance as we no longer have the method stack overhead and might even benefit from some caching. More importantly, this step opens up possibilities for future optimization. The recursion is usually eliminated by trying to reverse the order of the steps from the top-down approach. The observation to make here is that we only ever jump to the right. This means that if we start from the right of the array, every time we will query a position to our right, that position has already be determined as being GOOD or BAD. This means we don't need to recurse anymore, as we will always hit the memo table. Copy Copy Java enum Index { GOOD, BAD, UNKNOWN 3 4 public class Solution { 6 public boolean canJump(int[] nums) { 7 Index[] memo = new Index[nums.length]; for (int i = 0; i < memo.length; i++) { 8 memo[i] = Index.UNKNOWN; 9 10 11 memo[memo.length - 1] = Index.GOOD; 12 13 for (int i = nums.length - 2; i >= 0; i--) { int furthestJump = Math.min(i + nums[i], nums.length - 1); 14 15 for (int j = i + 1; $j \leftarrow furthestJump$; j++) { 16 if (memo[j] == Index.GOOD) { memo[i] = Index.GOOD; 17 break; 18

Approach 4: Greedy Once we have our code in the bottom-up state, we can make one final, important observation. From a given position, when we try to see if we can jump to a GOOD position, we only ever use one - the first one (see the break statement). In other words, the left-most one. If we keep track of this left-most GOOD position as a

9

U

for (int i = nums.length - 1; i >= 0; i--) {

• Space complexity : O(1). We are not using any extra memory.

if (i + nums[i] >= lastPos) {

public boolean canJump(int[] nums) { int lastPos = nums.length - 1;

lastPos = i;

return lastPos == 0;

return memo[0] == Index.GOOD;

0 and we need to decide if index 0 is GOOD. Since index 1 was determined to be GOOD, it is enough to jump there and then be sure we can eventually reach index 6. It does not matter that nums [0] is big enough to jump all the way to the last index. All we need is **one** way. Index 0 1 2 3 4 5 6

2

В

В

0

В

2

G

0

G

Сору

Next 🕖

Sort By -

Post

A Report

Conclusion The question left unanswered is how should one approach such a question in an interview scenario. I would say "it depends". The perfect solution is cleaner and shorter than all the other versions, but it might not be so

recursion) to bottom-up. Practicing similar problems will help bridge this gap.

= T(x+1) + T(x+1) $=2\cdot T(x+1)$ Now by induction, assume $T(x)=2^{n-x-1}$ and prove $T(x-1)=2^{n-(x-1)-1}$ $T(x-1) = 2 \cdot T(x)$ $= 2 \cdot 2^{n-x-1}$ $=2^{n-x-1+1}$ $=2^{n-(x-1)-1}$

Therefore, since we start from position 1, $T(1)=2^{n-2}$. Final complexity $O(2^{n-2})=O(2^n)$.

Thanks a lot. I hope you keep posting for other solutions **SHOW 5 REPLIES** Xulai_Cao 🛊 175 🗿 February 4, 2018 9:12 AM Sharing my solution as I didn't see it in the solutions

public boolean canJump(int[] nums) {

Read More

I made a video if anyone is having trouble understanding the solution (clickable link)

int len = nums.length:

class Solution {

SHOW 8 REPLIES

SHOW 5 REPLIES

86 A V C Share Reply

55 A V C Share Reply

https://youtu.be/2HnlGToCdCc

qinlei515 🛊 228 ② April 28, 2018 7:48 PM

As we already see, DP is totally unnecessary...

These solutions are a bit complicated. My greedy solution, very easy to understand: public boolean canJump(int[] nums) { if (nums.length == 0) throw new IllegalArgumentException("nums can't be empty"): Read More 31 A V Share Reply **SHOW 2 REPLIES**

next nums[i] elements to its right aiming to find a GOOD index. nums[i] can be at most n" That's correct but how many times you will going to do that? when you fail on an index you mark it as BAD so you only do that iteration once, because next time you visit a BAD index you return BAD directly without Read More 11 ^ C Share Reply **SHOW 3 REPLIES** arrayofchar # 46 ② September 12, 2018 10:08 AM

12 A V C Share Reply **SHOW 3 REPLIES** buckeyekarun 🛊 8 ② August 23, 2018 6:48 PM Assuming an input array of [99, 99, 99, 99],

8 A V C Share AReply **SHOW 7 REPLIES** cjavier70 🛊 13 🗿 January 3, 2019 11:53 PM

Average Rating: 4.92 (704 votes) Given an array of non-negative integers, you are initially positioned at the first index of the array.

6 9 6

Otherwise, that index is called a "bad index". The problem then reduces to whether or not index 0 is a This is a dynamic programming question. Usually, solving and fully understanding a dynamic programming 2. Optimize by using a memoization table (top-down³ dynamic programming) 3. Remove the need for recursion (bottom-up dynamic programming) All solutions presented below produce the correct result, but they differ in run time and memory the last. We start from the first position and jump to every index that is reachable. We repeat the process Copy int furthestJump = Math.min(position + nums[position], nums.length - 1); for (int nextPosition = position + 1; nextPosition <= furthestJump; nextPosition++) {</pre>

if (canJumpFromPosition(nextPosition, nums)) { return true; public boolean canJump(int[] nums) { return canJumpFromPosition(0, nums); 20 } One quick optimization we can do for the code above is to check the nextPosition from right to left. The theoretical worst case performance is the same, but in practice, for silly examples, the code might run faster. 0 1 2 3 5 6 Index 4

5 4 3 2 0 0 1 nums The first few steps of the backtracking algorithm for the example above are: 0 -> 4 -> 5 -> 4 -> 0 -> 3 -> 5 -> 3 -> 4 -> 5 -> etc. **Complexity Analysis** • Time complexity : $O(2^n)$. There are 2^n (upper bound) ways of jumping from the first position to the last, where n is the length of array nums. For a complete proof, please refer to Appendix A. • Space complexity : O(n). Recursion requires additional memory for the stack frames.

Steps 1. Initially, all elements of the memo table are UNKNOWN, except for the last one, which is (trivially) GOOD (it can reach itself) 2. Modify the backtracking algorithm such that the recursive step first checks if the index is known (GOOD) / BAD)

1

4

G

public boolean canJumpFromPosition(int position, int[] nums) {

return memo[position] == Index.GOOD ? true : false;

if (memo[position] != Index.UNKNOWN) {

Index

nums

memo

Java

3

6

8 9

10 11 12

22

23 24 25

26

27

19 20 21

22 23

24 25 }

altogether.

nums

memo

public class Solution {

Java

3

6

7 8

9

10 11 } }

straightforward to figure out.

References

O Previous

Comments: 1777

technique as a final thought in the interview.

}

Complexity Analysis

of array nums.

enum Index {

GOOD, BAD, UNKNOWN

return false;

of array nums.

of the memo table.

public boolean canJump(int[] nums) {

memo = new Index[nums.length];

for (int i = 0; i < memo.length; i++) {

public class Solution {

Index[] memo;

0

2

G

Complexity Analysis • Time complexity : $O(n^2)$. For every element in the array, say i , we are looking at the next nums[i]elements to its right aiming to find a GOOD index. nums[i] can be at most n, where n is the length

• Space complexity: O(2n) = O(n). First n originates from recursion. Second n comes from the usage

• Space complexity : O(n). This comes from the usage of the memo table.

(currPosition + nums[currPosition] >= leftmostGoodIndex). If we can reach a GOOD index, then our position is itself GOOD. Also, this new GOOD position will be the new leftmost GOOD index. Iteration continues until the beginning of the array. If first position is a GOOD index then we can reach the last index from the first position. To illustrate this scenario, we will use the diagram below, for input array nums = [9, 4, 2, 1, 0, 2, 0]. We write **G** for GOOD, **B** for BAD and **U** for UNKNOWN. Let's assume we have iterated all the way to position

separate variable, we can avoid searching for it in the array. Not only that, but we can stop using the array

Iterating right-to-left, for each position we check if there is a potential jump that reaches a GOOD index

• Time complexity : $O(n^2)$. For every element in the array, say i , we are looking at the next nums[i]

elements to its right aiming to find a GOOD index. nums[i] can be at most n, where n is the length

Complexity Analysis • Time complexity : O(n). We are doing a single pass through the **nums** array, hence n steps, where n is the length of array nums.

4

G

```
There are 2^n (upper bound) ways of jumping from the first position to the last, where n is the length of array
\overline{\mathsf{nums}} . We get this recursively. Let T(x) be the number of possible ways of jumping from position \mathsf{x} to
position n. T(n)=1 trivially. T(x)=\sum_{i=x+1}^n T(i) because from position x we can potentially jump to
all following positions i and then from there there are T(i) ways of continuing. Notice this is an upper
bound.
                                       T(x) = \sum_{i=x+1}^{n} T(i)
                                              =T(x+1)+\sum_{i=x+2}^n T(i)
```

The (recursive) backtracking is the easiest to figure out, so it is worth mentioning it verbally while warming

mention that there might be a dynamic programming solution and try to think how could you use a

up for the tougher challenge. It might be that your interviewer actually wants to see that solution, but if not,

memoization table. If you figure it out and the interviewer wants you to go for the top-down approach, it will

not generally be time to think of the bottom-up version, but I would always mention the advantages of this

Most people are stuck when converting from top-down Dynamic Programming (expressed naturally in

Appendix A - Complexity Analysis for Approach 1

Type comment here... (Markdown is supported)

1. https://en.wikipedia.org/wiki/Dynamic_programming

3. https://en.wikipedia.org/wiki/Top-down_and_bottom-up_design

2. https://en.wikipedia.org/wiki/Memoization

Rate this article: * * * * *

Read More 23 A V C Share Reply

terrible_whiteboard # 626 May 19, 2020 6:24 PM

silon_peter # 116 ② January 15, 2019 3:17 AM Dynamic Programming solutions O(n*n) time out in Python on the last test. No point of DP here. 24 A V Share Reply **SHOW 3 REPLIES** Bernoulli 25 October 13, 2018 12:06 PM

A rather straightforward one-pass solution if not nums: return False

Read More A Report I think the runtime complexity is n! and not 2^n for the recursive solution because position 0 -> 3 recursive calls maximum position 1 -> 2 recursive calls maximum

Read More

Why the Top-down approach time complexity is N^2?, while you are using memorization which mean you only visit each index once. you stated " For every element in the array, say i, we are looking at the

Best leetcode solution I've seen

(1 2 3 4 5 6 ... 17 18 >