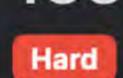
optimal solution can be built from the optimal solutions to smaller subproblems.

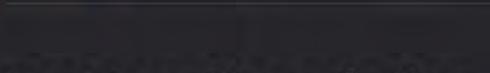
solution and understand how it aligns with the established approach:

variable mx holds the maximum job difficulty for these k jobs.

Let's walk through a small example to illustrate the solution approach.

4. State Transition: We update the DP array f at position [i] [j].





Problem Description

Array

Dynamic Programming

the i-th job, all jobs before it must be done. A job schedule has a "difficulty", which is calculated by summing up the daily difficulties over d days, and the difficulty of each day is given by the hardest (highest difficulty) job completed on that day.

In this problem, we're deciding how to schedule a list of jobs over d days. Jobs must be completed in a specific order — to complete

The goal is to find the job schedule with the minimum total difficulty possible while ensuring that at least one job is finished each day. If you're given a set of jobs, each with its own difficulty level (stored in the array jobDifficulty), and a certain number of days d, your task is to find this minimum difficulty. If it's not possible to schedule at least one job per day, the function should return -1.

painting d sets of houses. Each day, you can paint a continuous block of houses, and the difficulty of the day is the hardest house you painted. Your goal is to minimize your total effort over the deadline without taking any days off.

To illustrate, imagine you're painting houses, and each house has a different level of effort required. You have a deadline to finish

The intuitive approach to this kind of problem involves understanding that it's a classic case for dynamic programming, where the

Intuition

We create a 2D array f where f[i][j] represents the minimum difficulty to complete the first i jobs in j days. Initially, we know that f[0][0] = 0 because no difficulty exists if no jobs are done in 0 days. For all other values, we initialize f[i][j] to infinity (inf) since

we have not yet computed the minimum difficulty for those scenarios. As we attempt to solve for f[i][j], we encounter a choice: which jobs to schedule on the j-th day? We could select any sequence of jobs ending with the i-th job to be done on the last day. This selection defines the difficulty of the j-th day by the hardest job in

that sequence. The remaining jobs would have been completed in the previous j-1 days. So, for each day j and job i, we iterate backward through the job list, calculating the difficulty for doing the last k jobs on day j, where k ranges from i down to 1. We determine the max job difficulty mx for the last k jobs and add it to the previously computed

minimum difficulty needed to finish the first k-1 jobs in j-1 days. This gives us the total difficulty if we were to complete the last k

jobs on the current day j. We keep track of the minimum value for f[i][j] as we iterate through all possible k.

The final answer to our scheduling problem is represented by f[n][d], which tells us the minimum total difficulty to finish all n jobs within d days, based on how the jobs were divided across the days.

Solution Approach The problem is solved using a dynamic programming algorithm. Let's break down the implementation provided in the reference

1. Initialization: We create a 2D list f that will serve as our DP table with n + 1 rows and d + 1 columns. Each element is initially set to infinity (inf). The table is used to store the minimum difficulty of completing i jobs in j days. We also know that f[0][0]

should be 0 because no difficulty exists when no jobs are done in 0 days. 2. DP Table Population: We use two nested loops to populate this table. The outer loop iterates over the range [1, n + 1) to

to schedule jobs in more days than we have jobs or days allocated. 3. Finding the Minimum Difficulty: Within the inner loop, we introduce another loop iterating from i down to 1 (backwards). This iteration is critical; for each job i and day j, it checks all possibilities of completing some last k jobs on the current day j. The

represent the jobs. The inner loop iterates over the range [1, min(d + 1, i + 1)) to represent the days, ensuring we don't try

- 4. State Transition: For each k, we calculate a candidate minimum difficulty by taking the previously computed minimum difficulty for k-1 jobs done in j-1 days (f[k - 1][j - 1]) and adding the difficulty of completing the current k jobs on the current day, represented by mx. This reflects the state transition equation: 1 f[i][j] = min(f[i][j], f[k-1][j-1] + mx)
- Here, f[i][j] holds the minimum difficulty for i jobs done in j days, and we update it only if we find a lower difficulty than the current stored value. 5. Returning the Result: After populating the DP table, we look at f[n] [d] to see the result. If we find that it's still inf, it means that

it's not possible to schedule the jobs within d days, and we should return -1. Otherwise, f[n] [d] contains the minimum difficulty

```
job schedule.
The key algorithms used here are dynamic programming for solving the optimized subproblems and iterating in reverse order to
```

meeting all the constraints of the problem. Example Walkthrough

By carefully updating our DP table and checking all possibilities, we guarantee that we find the minimum difficulty job schedule while

identify the possible subsets for achieving daily goals. The use of a 2D DP table for storing intermediate results is a common practice

Suppose we have the following job difficulties and days: 1 jobDifficulty = [7, 1, 4, 5] 2 d = 2

1. Initialization: We create a 2D list f with 5 (n + 1) rows (representing 0 to 4 jobs) and 3 (d + 1) columns (representing 0 to 2 days).

Step-by-Step Solution:

1 f = [[0, inf, inf],[inf, inf, inf], [inf, inf, inf], [inf, inf, inf],

This results in a 5×3 matrix initialized to infinity, except for f[0][0] which is 0.

We want to schedule these 4 jobs over 2 days such that the total difficulty of the schedule is minimized.

in dynamic programming problems to avoid recalculating solutions to subproblems multiple times.

```
2. DP Table Population: We iterate over jobs (i from 1 to 4) and for each job, over days (j from 1 up to the current job number i
```

without exceeding day d).

Moving to i=2, we need to check:

If we do job 1 and 2 on day 1, then:

[inf, inf, inf]]

 For i=1 and j=1, the only option to schedule the first job on the first day means: 1 f[1][1] = 7 (which is jobDifficulty[0])

3. Finding the Minimum Difficulty: We perform an inner backward iteration from 1 down to 1 for each job/day combination.

• If we do job 2 on day 1 and job 1 on day 0 which is not possible, so we ignore it. Now for i=2 and j=2, we need to check: If we do job 1 on day 1 and job 2 on day 2, then:

k jobs on the current day j and adding it to the minimum difficulty of scheduling the first i - k jobs on the previous days j - 1.

5. Returning the Result: After populating the table, we check f[4][2] for our answer. If it's inf, then it's impossible to schedule,

Carrying on with this process till we complete all the jobs, we keep updating f based on the maximum difficulty of scheduling the last

but in this case, suppose it is 15, that would be our minimum difficulty to schedule all jobs over 2 days as per our example. Finally, the DP table f might look something like this after being populated (with hypothetical values for illustration):

1 f[2][1] = max(jobDifficulty[0], jobDifficulty[1]) = max(7, 1) = 7

1 f[2][2] is min(inf, f[1][1] + max(jobDifficulty[1])) = min(inf, 7 + 1) = 8

And the result is f[4][2] = 15, which is the minimum total difficulty to schedule all 4 jobs over 2 days.

Only consider scheduling up to the minimum of days or jobs done so far

return -1 if dp_table[num_jobs][days] == float('inf') else dp_table[num_jobs][days]

Try to end the j-th day with each possible last job

class Solution: def minDifficulty(self, job_difficulty: List[int], days: int) -> int: num_jobs = len(job_difficulty) # Initialize DP table with infinity representing impossible scenarios. # dp_table[i][j] will hold the minimum difficulty of scheduling the first i jobs in j days

max_difficulty_on_last_day = max(max_difficulty_on_last_day, job_difficulty[k - 1])

 $dp_table[i][j] = min(dp_table[i][j], dp_table[k - 1][j - 1] + max_difficulty_on_last_day)$

If the difficulty of scheduling all jobs in d days is infinite, it means it is not possible, thus return -1.

final int MAX_DIFFICULTY = Integer.MAX_VALUE / 2; // Define a high number to represent an impossible situation

Update the DP table by adding the difficulty of the last job to the optimal difficulty of previous jobs in j-

```
dp_table = [[float('inf')] * (days + 1) for _ in range(num_jobs + 1)]
 8
           dp table[0][0] = 0 # Base case: 0 jobs in 0 days has 0 difficulty
 9
10
11
           # Loop through each job
```

12

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25

Java Solution

class Solution {

Python Solution

from typing import List

for i in range(1, num_jobs + 1):

for j in range(1, min(days + 1, i + 1)):

max_difficulty_on_last_day = 0

Otherwise, return the calculated difficulty

public int minDifficulty(int[] jobDifficulty, int d) {

int minDifficulty(vector<int>& jobDifficulty, int d) {

int numJobs = jobDifficulty.size();

memset(minDiff, 0x3f, sizeof(minDiff));

int maxDifficulty = 0;

for (int k = i; k; ---k) {

int minDiff[numJobs + 1][d + 1];

for k in range(i, 0, -1):

1 f = [[0, inf, inf],

[7, inf, inf],

[10, 11, inf],

[15, 12, 15]]

[7, 8, inf],

```
int n = jobDifficulty.length; // The total number of jobs
           int[][] dp = new int[n + 1][d + 1]; // Dynamic programming table
 6
           // Initialize all dp table values with MAX_DIFFICULTY, except for the starting point
           for (int[] row : dp) {
 8
               Arrays.fill(row, MAX_DIFFICULTY);
 9
10
11
           dp[0][0] = 0; // Base case: 0 jobs on day 0 has difficulty 0
12
13
           // Fill in the dynamic programming table
           for (int i = 1; i <= n; ++i) { // For each job
14
               for (int j = 1; j \le Math.min(d, i); ++j) { // For each day, until the min of current job index and days
15
16
                   int maxDifficulty = 0; // To keep track of the maximum difficulty of jobs for the current day
17
                   // Iterate through the previous jobs to find the minimum difficulty
                   for (int k = i; k > 0; --k) {
18
                       maxDifficulty = Math.max(maxDifficulty, jobDifficulty[k - 1]); // Find max difficulty among jobs
19
                       // Update the dp table for the minimum difficulty after completing current job on day j
20
                       dp[i][j] = Math.min(dp[i][j], dp[k - 1][j - 1] + maxDifficulty);
21
22
23
24
           // If the final value is unmodified from the MAX_DIFFICULTY, it means it's not possible to schedule jobs within d days
25
           return dp[n][d] >= MAX_DIFFICULTY ? -1 : dp[n][d];
26
27
28 }
29
C++ Solution
```

minDiff[0][0] = 0;12 13 // Iterate through all jobs for (int i = 1; i <= numJobs; ++i) { 14 // For each job, iterate through all possible days, but never more than i days 15 for (int day = 1; day \leftarrow min(d, i); ++day) {

1 class Solution {

2 public:

6

9

10

11

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18

19

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16

17

18

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21

22

23

24

26

25 }

```
22
                        maxDifficulty = max(maxDifficulty, jobDifficulty[k - 1]);
23
24
                        // Recurrence relation:
                       // minDiff for job i on day 'day' is the minimum of its current value or
25
26
                       // the sum of max difficulty encountered today and
27
                       // the best we could have done through job k-1 on day 'day-1'
28
                        minDiff[i][day] = min(minDiff[i][day], minDiff[k - 1][day - 1] + maxDifficulty);
29
30
31
32
33
           // Check if it's possible to schedule jobs within d days. If not, return -1
34
           return minDiff[numJobs][d] == 0x3f3f3f3f ? -1 : minDiff[numJobs][d];
35
36 };
37
Typescript Solution
   function minDifficulty(jobDifficulty: number[], days: number): number {
       const numOfJobs = jobDifficulty.length; // number of jobs
       const INF = 1 << 30; // represents an infinite difficulty, used as an initial value</pre>
       const dp: number[][] = new Array(numOfJobs + 1)
           .fill(0)
           .map(() => new Array(days + 1).fill(INF)); // initialize the dp array
       dp[0][0] = 0; // base case: 0 jobs done in 0 days has a difficulty of 0
 8
       // Populate the dp table
9
       for (let i = 1; i <= numOfJobs; ++i) {</pre>
10
           for (let j = 1; j \le Math.min(days, i); ++j) { // Each job can only be scheduled if we have enough days
11
                let maxDifficulty = 0; // tracks the maximum difficulty of the current job set
12
13
               // Loop to find the minimum difficulty if the current job ends the current day
14
15
               for (let k = i; k > 0; --k) {
```

maxDifficulty = Math.max(maxDifficulty, jobDifficulty[k - 1]);

// If it's still INF, it means it's impossible to schedule all jobs, hence return -1

// Return the minimum difficulty to complete all jobs in the given days

return dp[numOfJobs][days] < INF ? dp[numOfJobs][days] : -1;

dp[i][j] = Math.min(dp[i][j], dp[k - 1][j - 1] + maxDifficulty);

// Create array to store the minimum difficulty on day j having completed job i

// Looking for the last job on the previous day to minimize today's difficulty

// Update the max difficulty for today's job set (k-1 to i)

// Initialize the array with high initial values, except for start state

```
Time and Space Complexity
The time complexity of the provided code is 0(n^2 * d). This is because there are three nested loops where the outermost loop runs
for n iterations (jobs), the middle loop runs for at most d iterations (days), and the innermost loop also runs for at most n iterations,
but on average will run fewer times as k decreases. However, in the worst case, the inner loop still contributes n operations per
iteration of the middle loop. The product of these gives the time complexity of 0(n^2 * d).
```

The space complexity of the code is 0(n * d). This arises from the use of a 2D array f with dimensions n + 1 by d + 1. Since the only other variables are constant size integers and not dependent on n or d, f dominates the space complexity.