

# Dynamic Programming 3

- Priyansh Agarwal

# Removal game (SES)

State

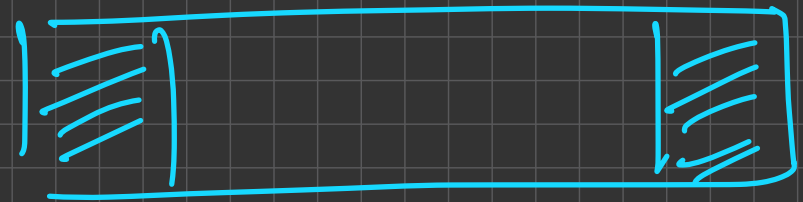
$dp[i][j] = \max$  sum the current  
player can get from subarray (i to j)

Transition

$$dp[i][j] \begin{cases} arr[i] + sum[i+1][j] - dp[i+1][j] \\ arr[j] + sum[i][j-1] - dp[i][j-1] \end{cases}$$

B.C =  $dp[i][i] = arr[i]$  , f.s =  $dp[0][n-1]$

dp(i,j)



↳ (max score that 1st player  
can get from (i to j))

dp(0)(n-1)

arr(0) + dp(1)(n-1)

max  
arr(n-1) + dp(0)(n-2)

dp(1)(n-1)

min  
arr(1) + dp(2)(n-1)  
arr(n-1) + dp(1)(n-2)

$(0, 5) \rightarrow$  1st player

length of array = 6

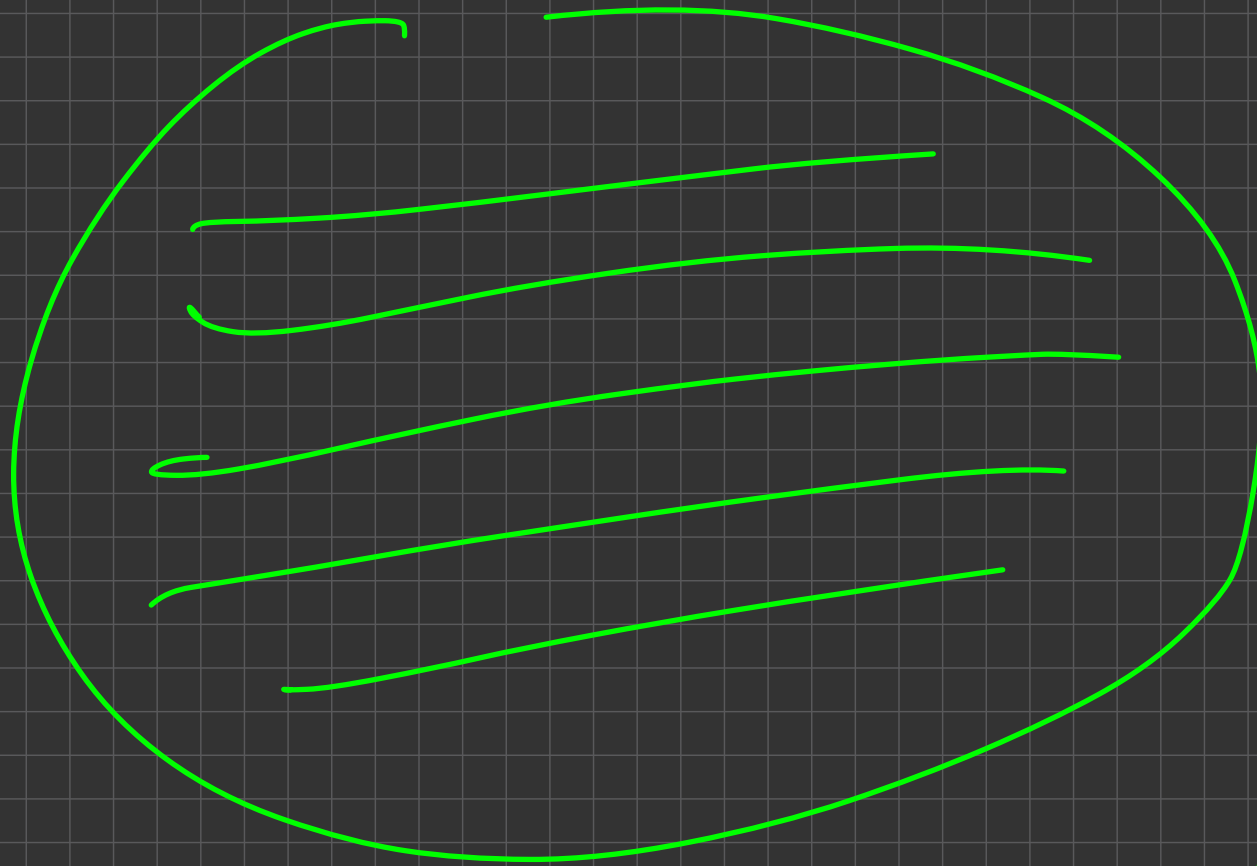
even — 1st player

odd — 2nd player

length of array = 9

odd — 1st player

even — 2nd player



A

B

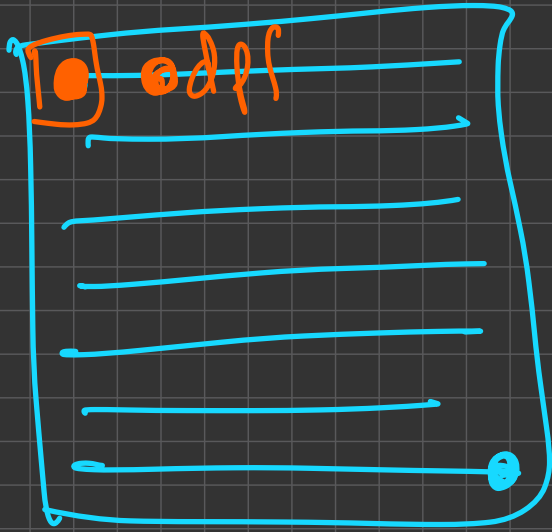
- Answer Construction ✓
- 1 Hard DP problem ✓
- Space optimisation ✓
- State optimisation
- Transition optimisation ✓

# $\propto$ no. of ways Answer Construction

→ checking for possibility

DP problem  
minimize / maximize  
value

- Grid problem: Find the actual path with the minimum sum.
- ~~Minimizing the problem. Start from the source and find the best choice at every state.~~
- At every state we are making some optimal choice.
  - If we store this choice, we can be sure that if we are at any state we know what is the best choice.
  - Start from the state that contains your final subproblem and keep making the best choice (which was already stored) until you reach the end.



minimize sum of values on  
the path

find out the exact path

$dp(i, j) = \text{min possible sum path from } (i, j) \text{ to } \underline{(n-1, m-1)}$

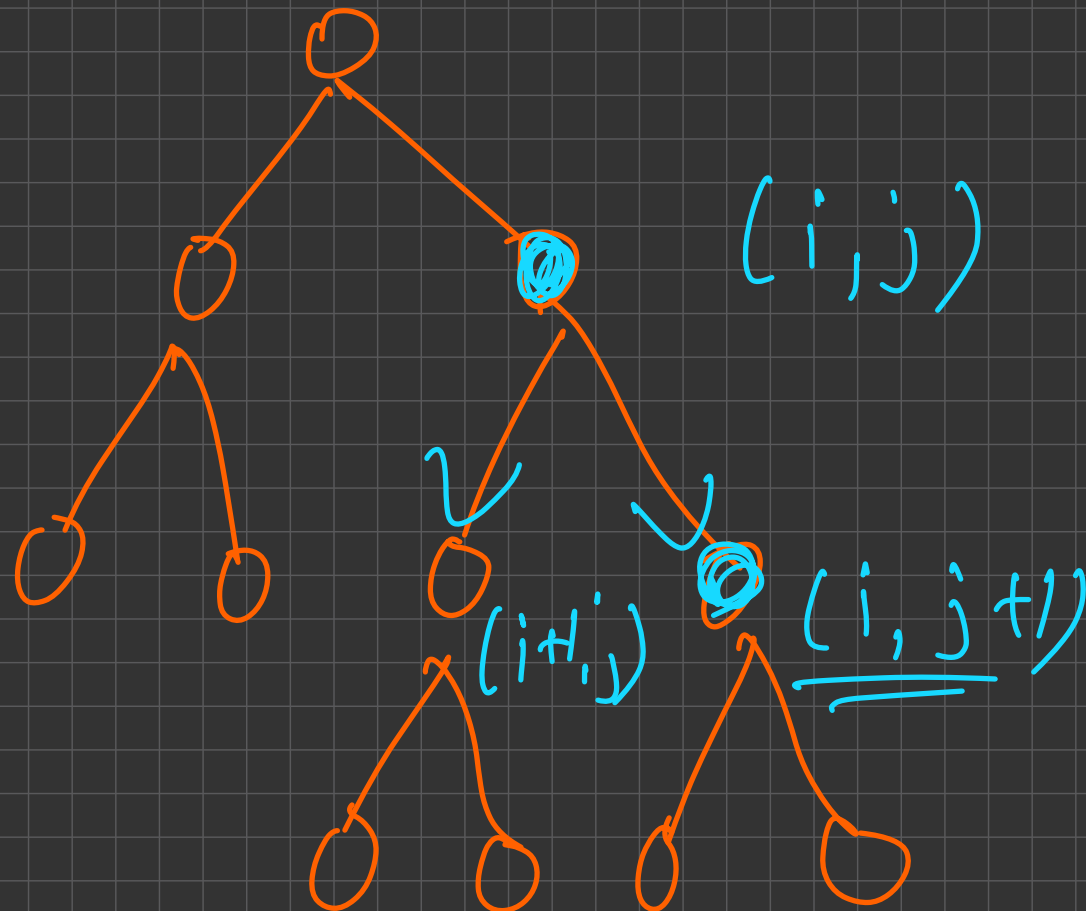
$dp(0, 1)$   
 $dp(1, 0)$

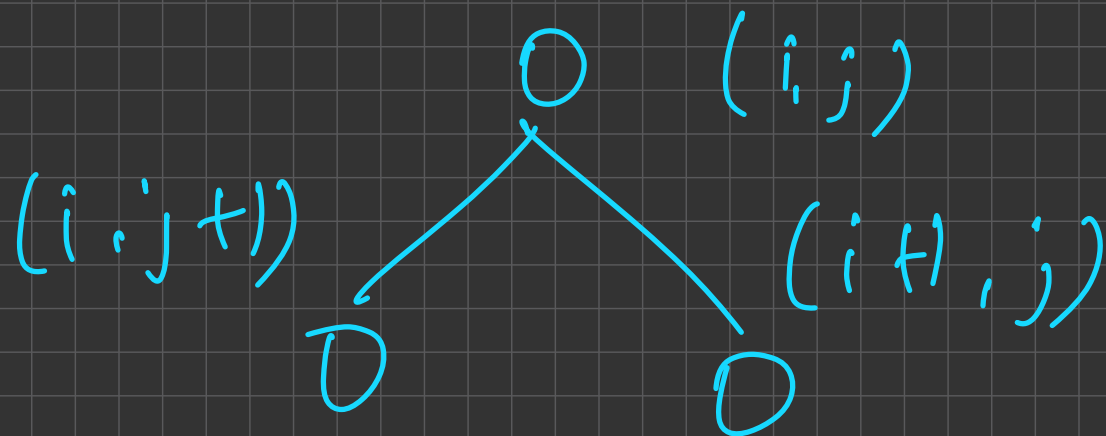
$(0, 0)$

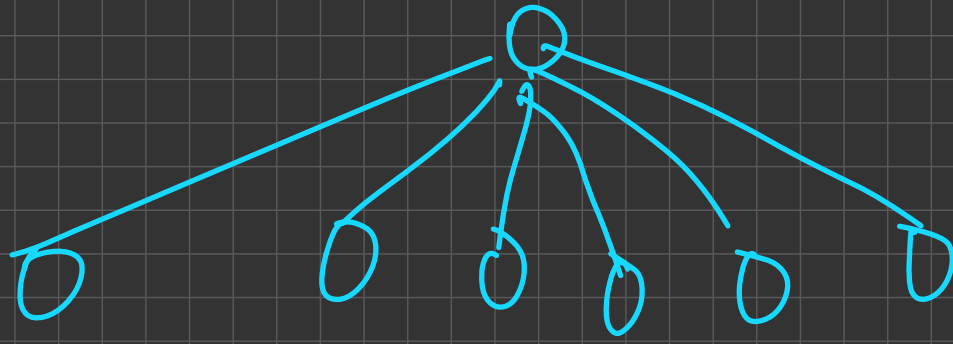
$(i, j) \rightarrow dp(i+1, j)$

$\searrow dp(i, j+1)$







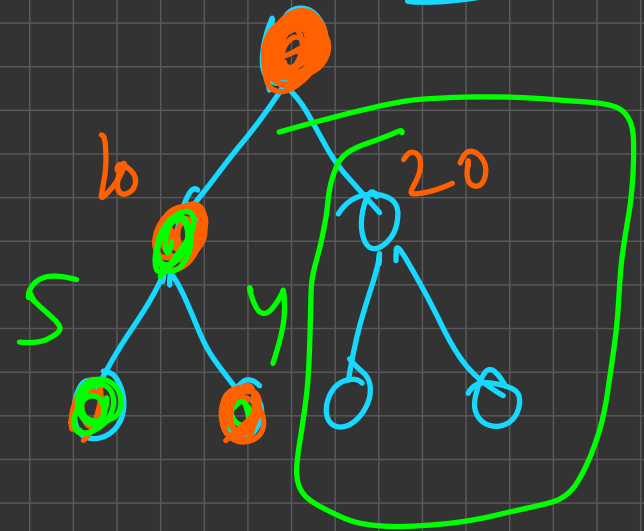


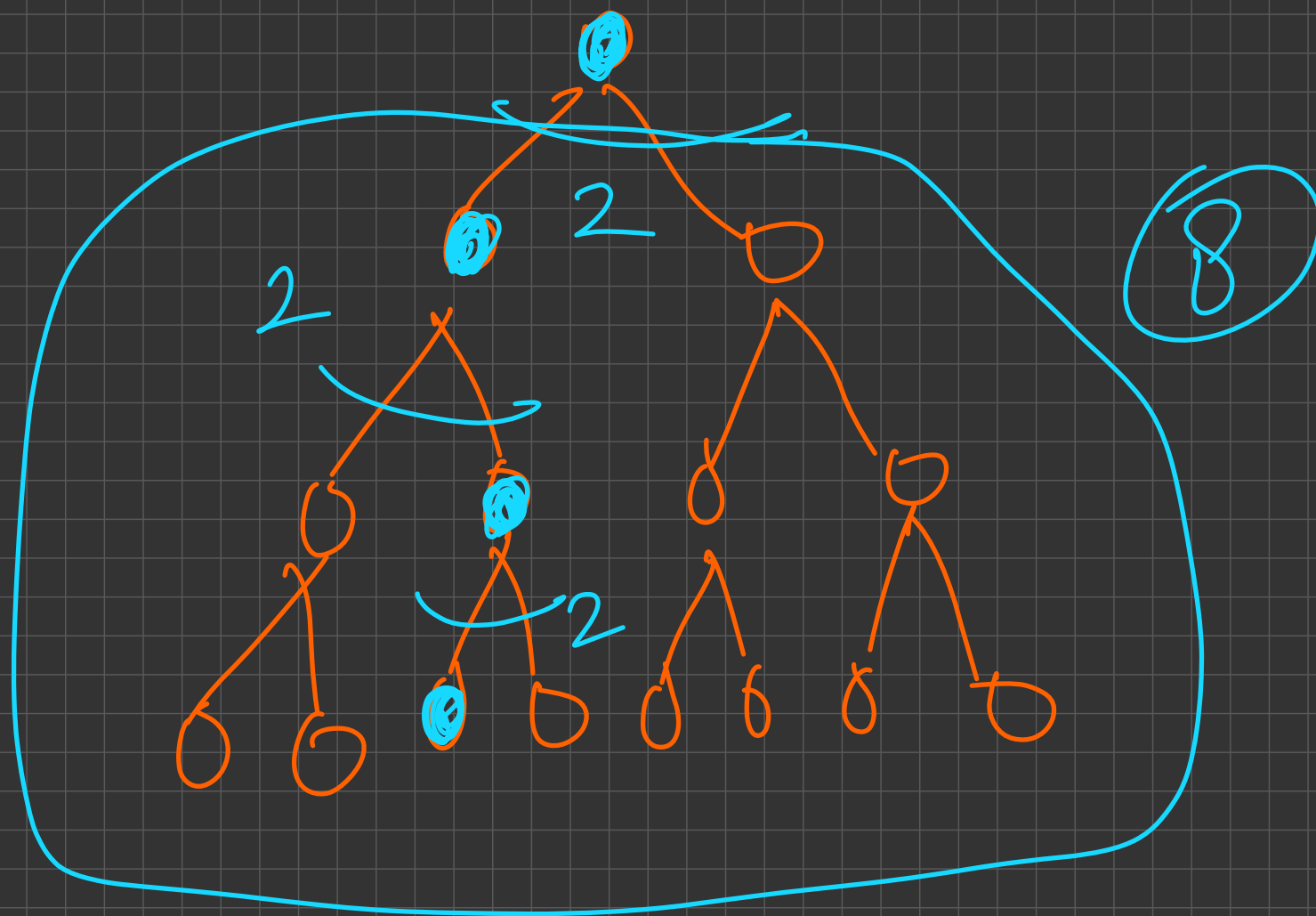
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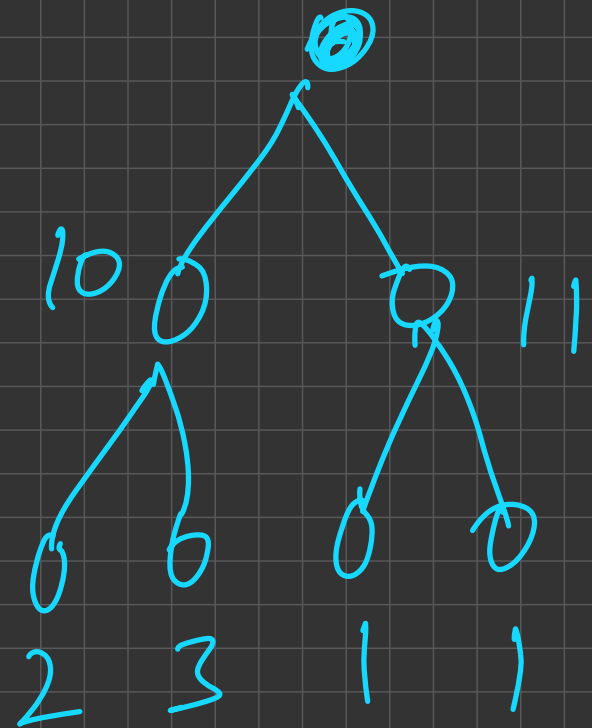
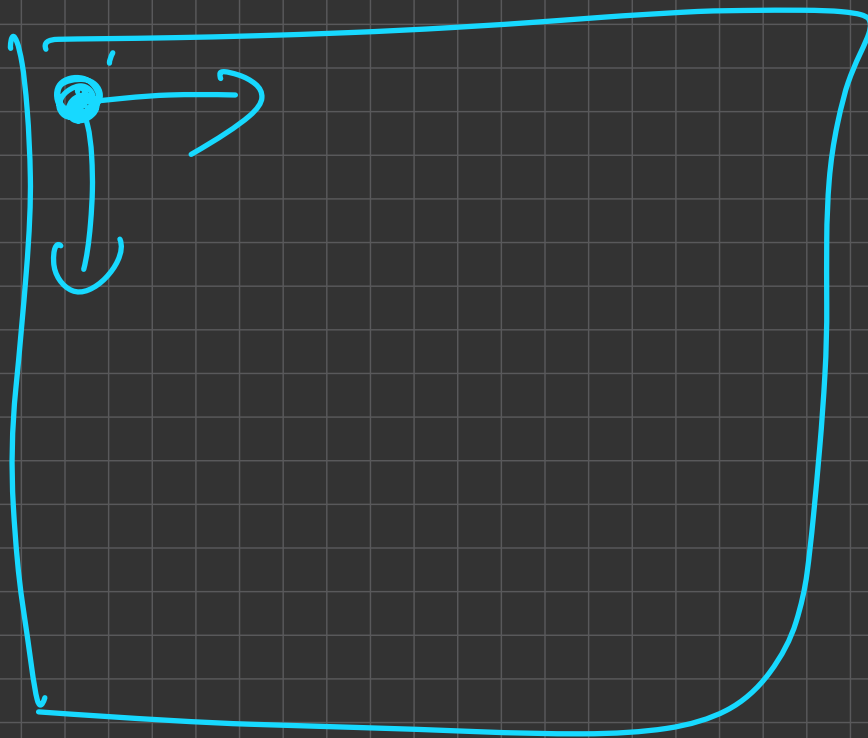
10 choices

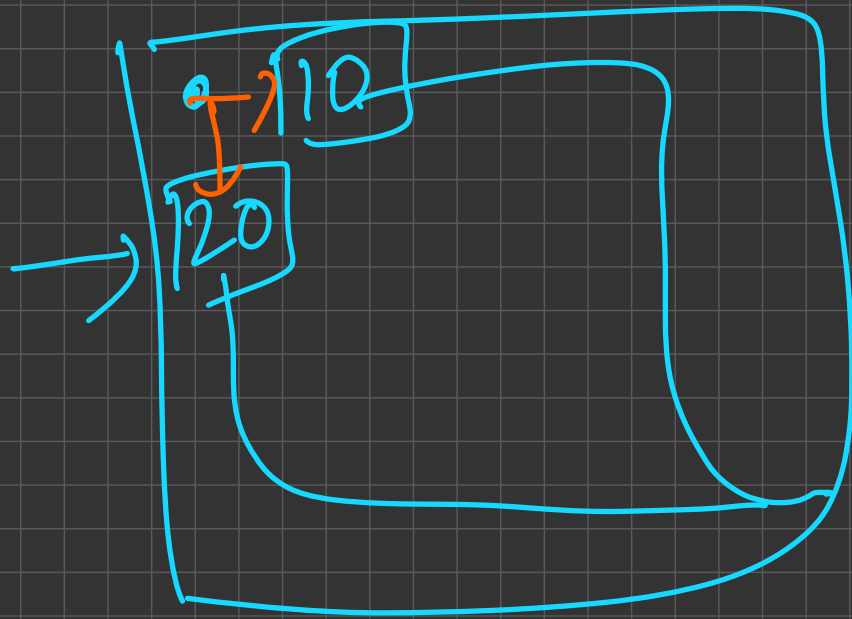
Grid problem  
=

100 cells







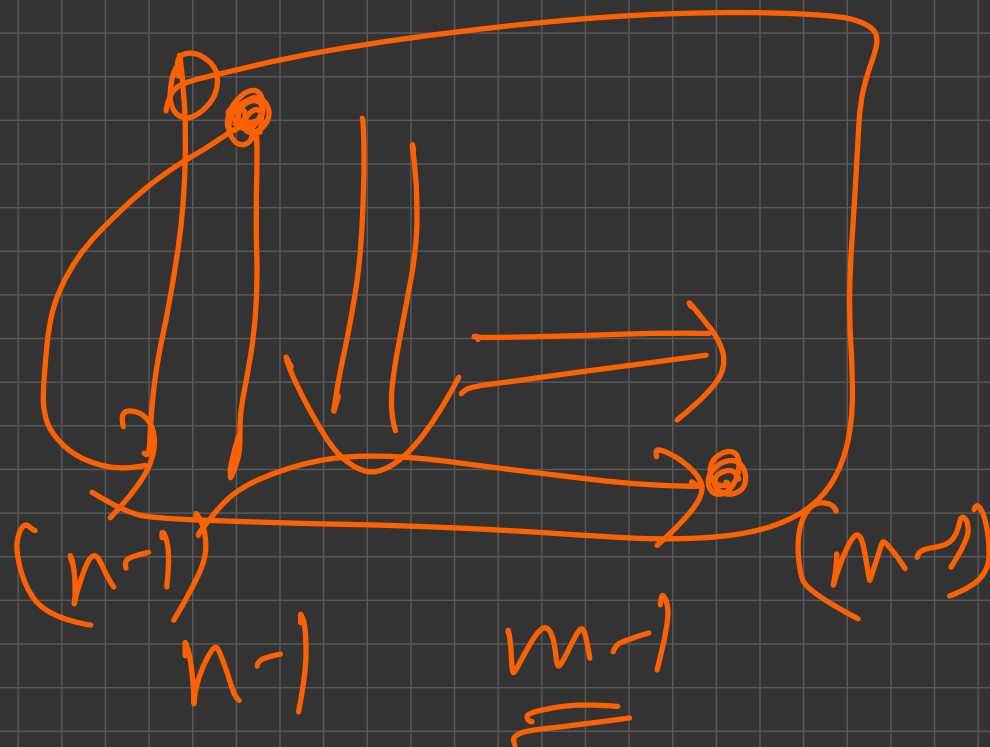


10

$2^{10}$

$2 \cdot 10$

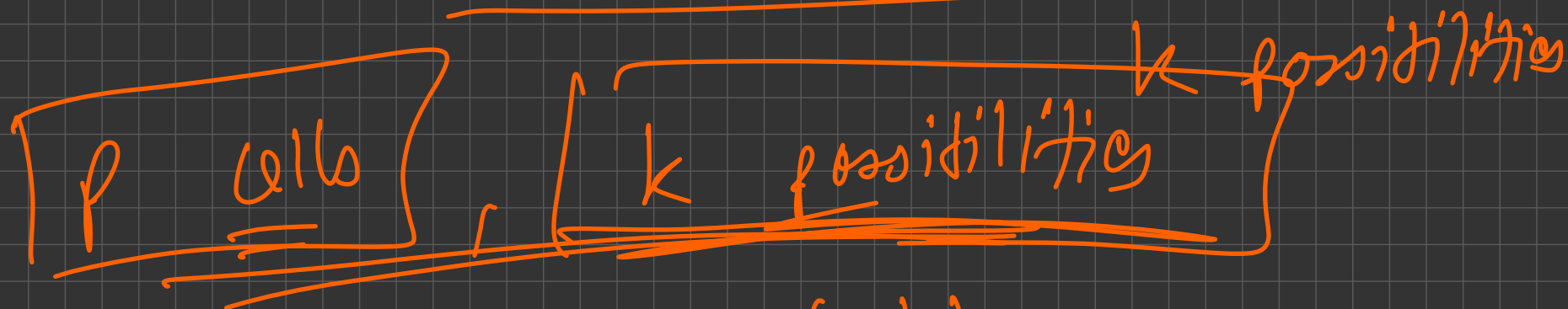
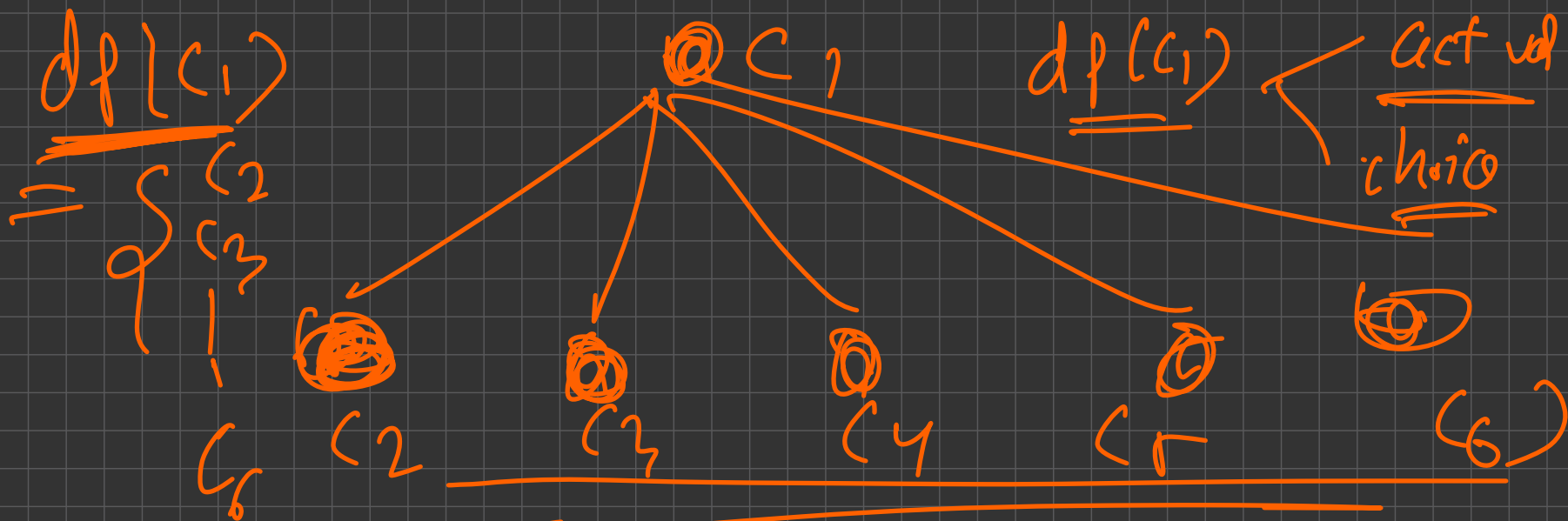
$dp(i,j) = \text{min sum path from } (i,j)$   
to  $(n-1, m-1)$



$$\underline{\underline{(n+m-2)}}$$

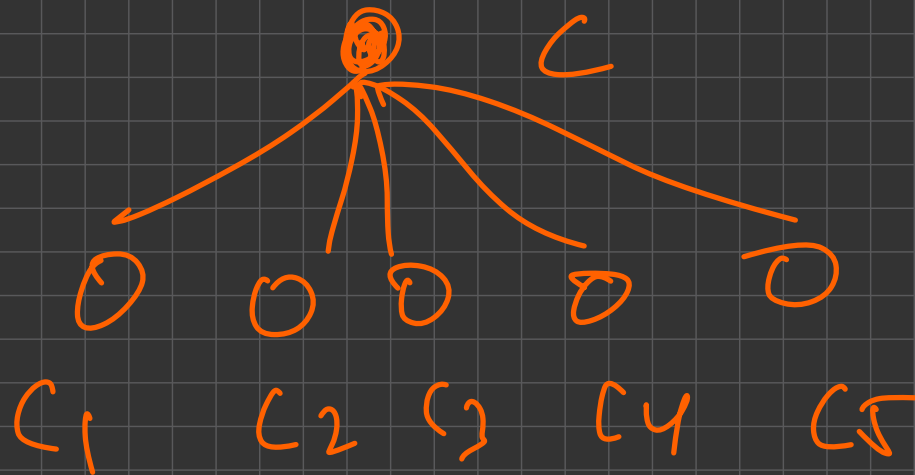
$$\underline{\underline{(n+m-1) \cdot 2}}$$

$$O(2 \cdot (n+m-1)) = \underline{\underline{O(n+m-1)}}$$



$\underline{O(p \cdot k)}$





$$df(c) = v + \min \{ \begin{matrix} C1 \\ C2 \\ C3 \\ C4 \\ C5 \end{matrix} \}$$

actual

$df(c) < \underline{\text{choice}} \quad \underline{C4}$

P

P.K  
OP

# Answer Construction - Grid Problem

```
int n = 3, m = 3;
vector<vector<int>> grid(3, vector<int>(3));
vector<vector<pair<int, int>>> dp(n, vector<pair<int, int>>(m, {-1, 0}));
// 0 -> take a down direction
// 1 -> take a right direction
int f(int i, int j){
    if(i == n || j == m)
        return 1e9;
    if(i == n - 1 && j == m - 1)
        return grid[n - 1][m - 1];
    if(dp[i][j].first != -1)
        return dp[i][j].first;

    int ans1 = f(i + 1, j);
    int ans2 = f(i, j + 1);
    if(ans1 < ans2){
        dp[i][j].second = 0;
    }else{
        dp[i][j].second = 1;
    }
    return dp[i][j].first = grid[i][j] + min(ans1, ans2);
}
```

act val , choice

$dp(i|j).first \rightarrow$  val  
 $dp(i|j).sec \rightarrow$  choice

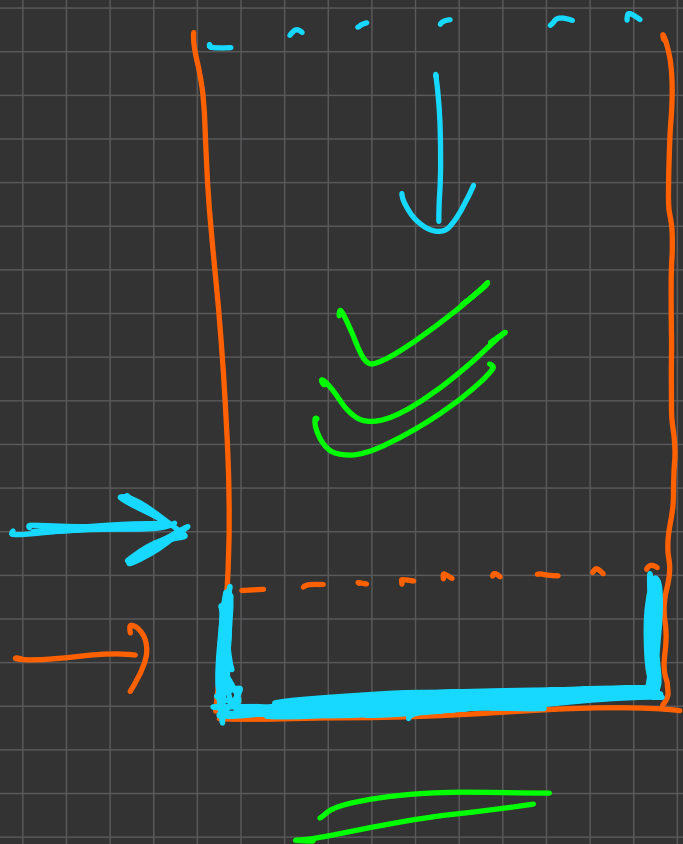
# Answer Construction - Grid Problem

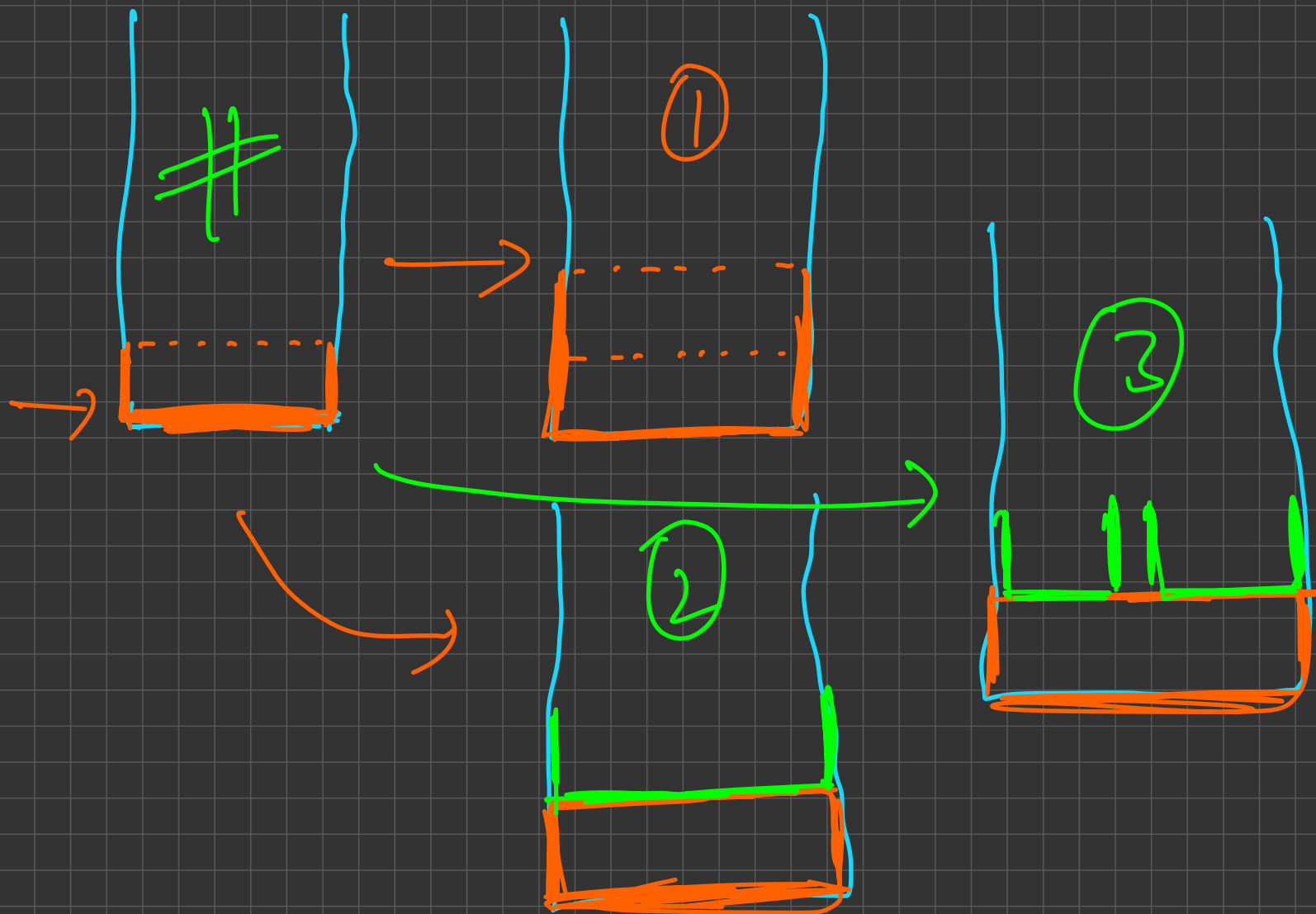
```
void solve(){  
    for(int i = 0; i < 3; i++){  
        for(int j = 0; j < 3; j++){  
            cin >> grid[i][j];  
        }  
    }  
    cout << f(0, 0) << nline;  
    pair<int, int> current = {0, 0};  
    while(current != mp(n - 1, m - 1)){  
        cout << current.first << " " << current.second << nline;  
        if(dp[current.first][current.second].second == 0)  
            current.first++;  
        else  
            current.second++;  
    }  
    cout << current.first << " " << current.second << nline;  
}
```

evaluate

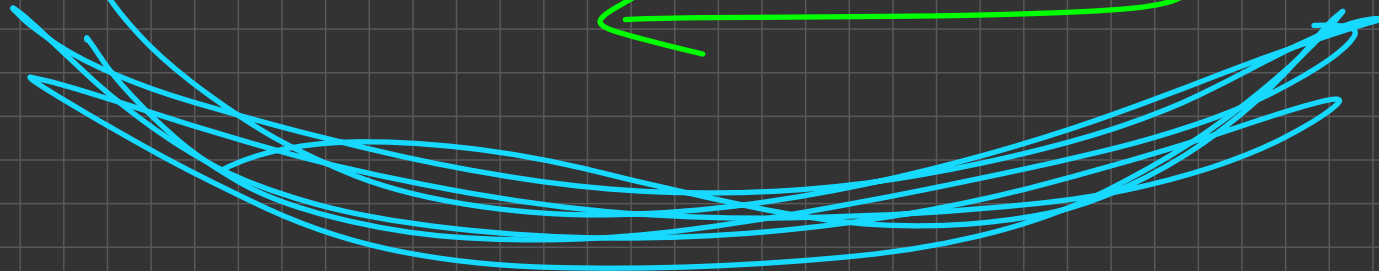
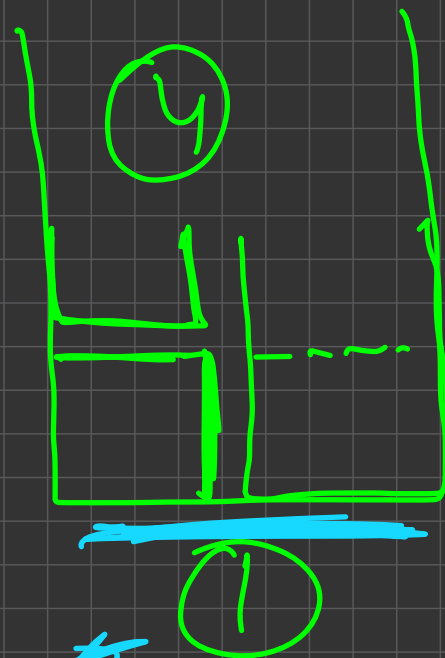
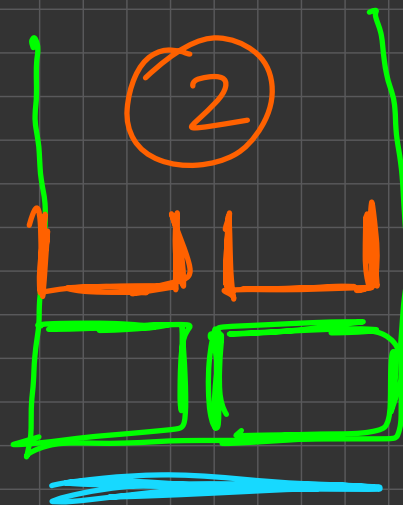
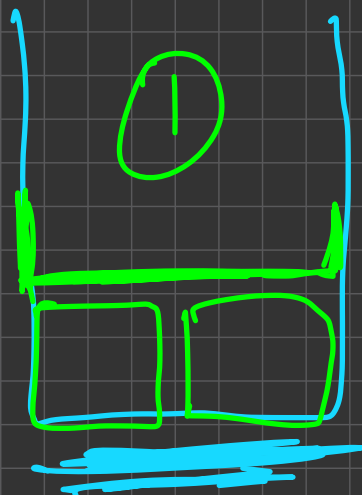
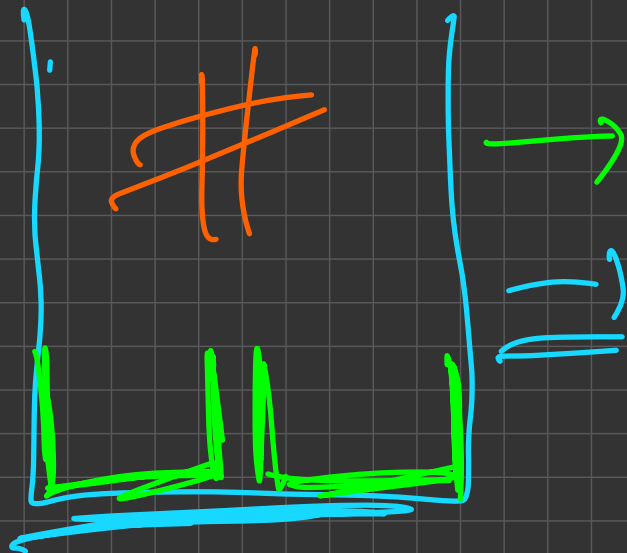
back track

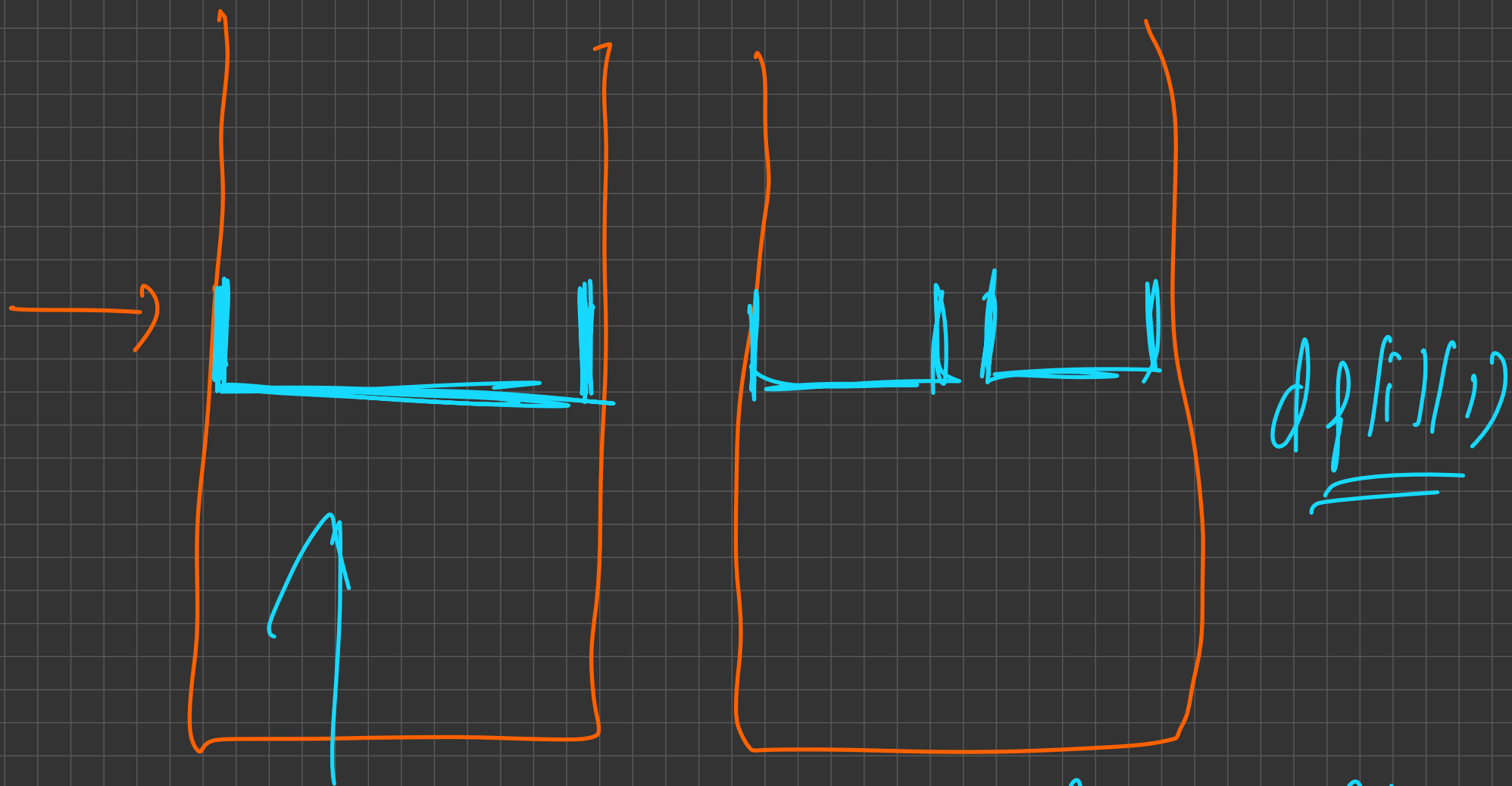
## Problem 1: [Link](#)





0





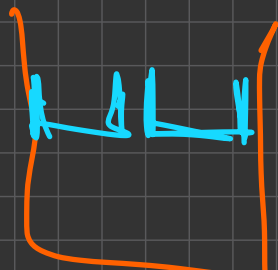
$dp[i][0]$  = no. of ways to fill up the tower from  $(i+1)^{th}$  index such that you have a  $1 \times 2$  cell extending



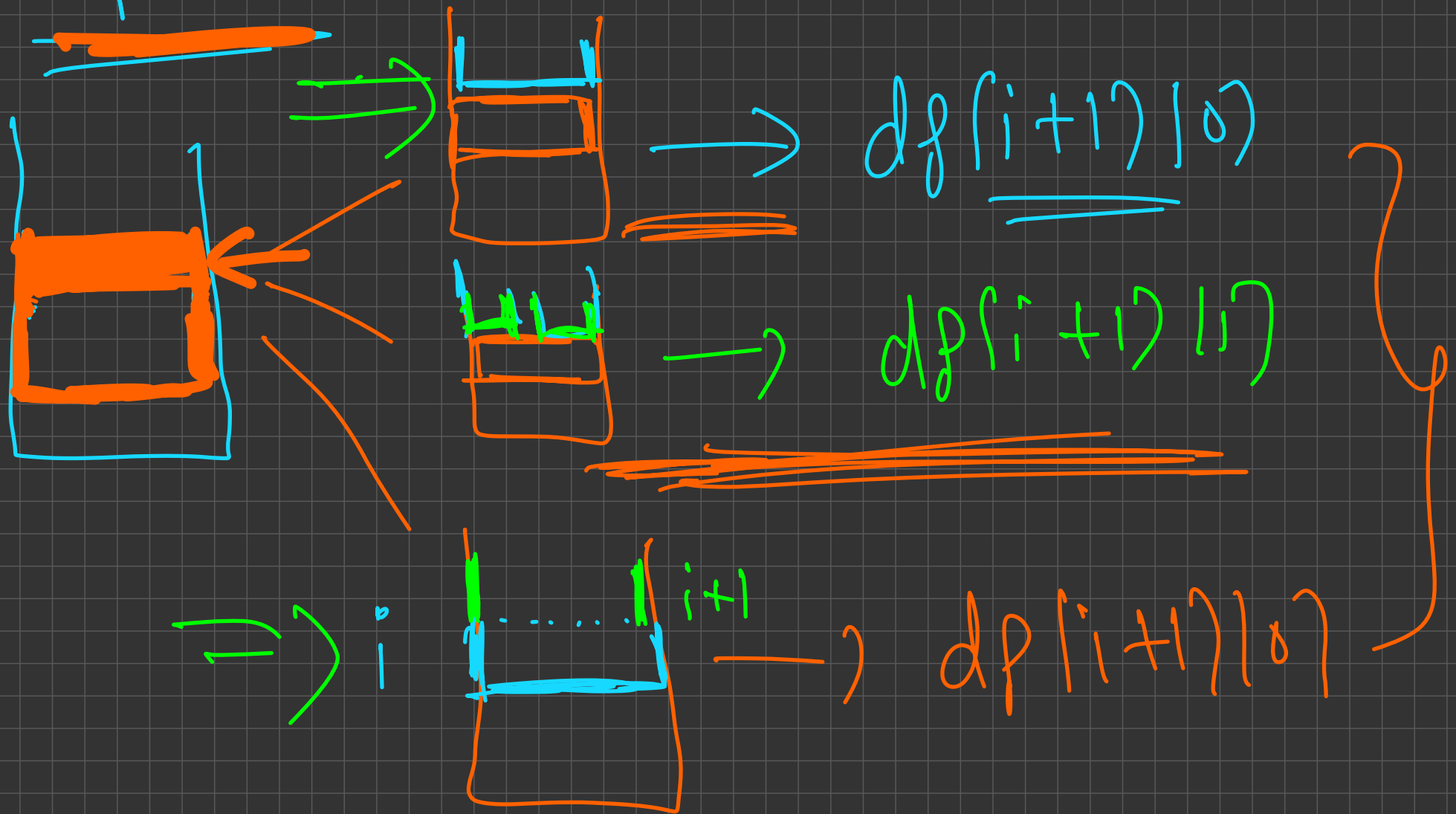
$dp(i|0) =$  no of ways to fill the tower from  $(i+1)$ th row to the top  
st you have a  $7 \times 2$  cell entirely



$dp(i|1) =$  no of ways to fill the tower from  $(i+1)$ th row to the top  
st you have 2  $(1 \times 1)$  cells entirely

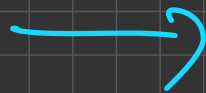
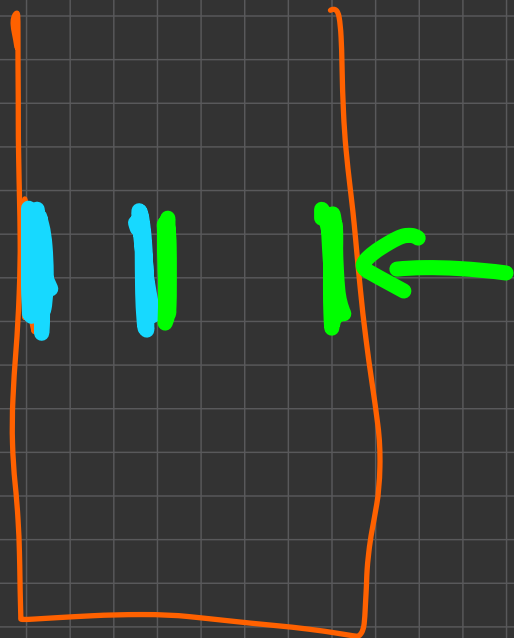


dp(i) | 0

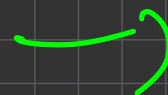


$$\underline{\underline{\text{dp}(i) | 0} = 2 \text{dp}(i+1) | 0 + \text{dp}(i+1) | 1}$$

$dp[i][0]$



$dp[i+1][0]$



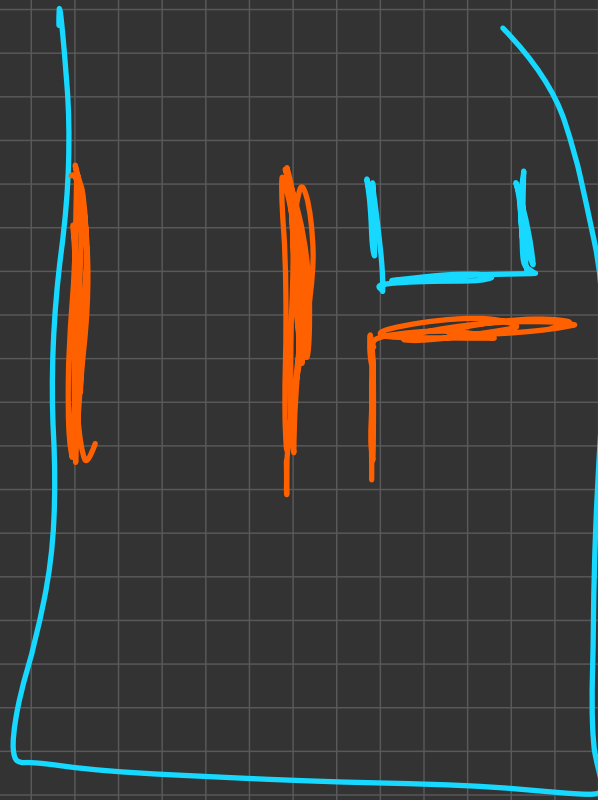
$dp[i+1][1]$



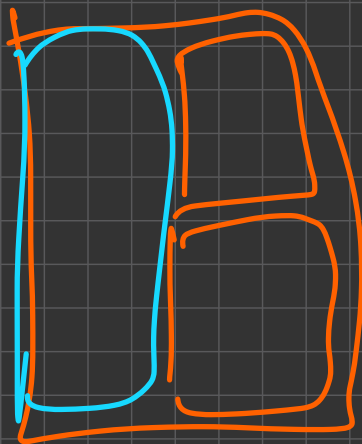
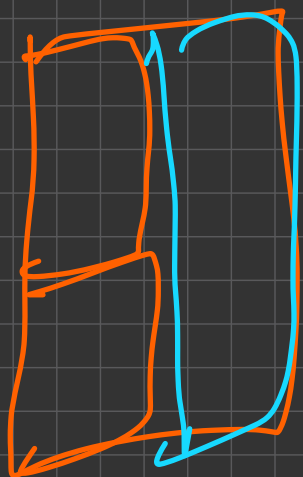
$dp[i+1][2]$



$\rightarrow dp(i+1)(1)$



$\leftarrow dp(i+1)(1)$



$$dp[i][1] = dp[i+1][0]$$

$$+ 4 \cdot \underline{\underline{dp[i+1][1]}}$$

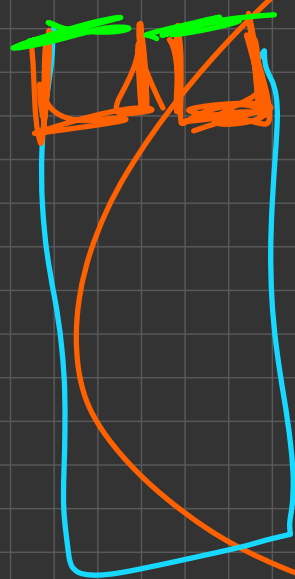
$$\underline{\underline{O(1)}}$$

$$\underline{\underline{O(1)}} \quad \underline{\underline{S.C}}$$

$$\underline{\underline{O(n)}} \quad \underline{\underline{T.C}}$$

$$\underline{dp(i)(0)} = 2 \cdot dp(i+1)(0) + dp(i+1)(1)$$

$$\underline{dp(i)(1)} = 4 \cdot dp(i+1)(1) + dp(i+1)(0)$$



$$dp(n-1)(0) = 1$$

$$dp(n-1)(1) = 1$$

B.C

$dp(i, 0)$   $\swarrow$   $dp(i+1, 0)$   
 $\searrow$   $dp(i+1, 1)$

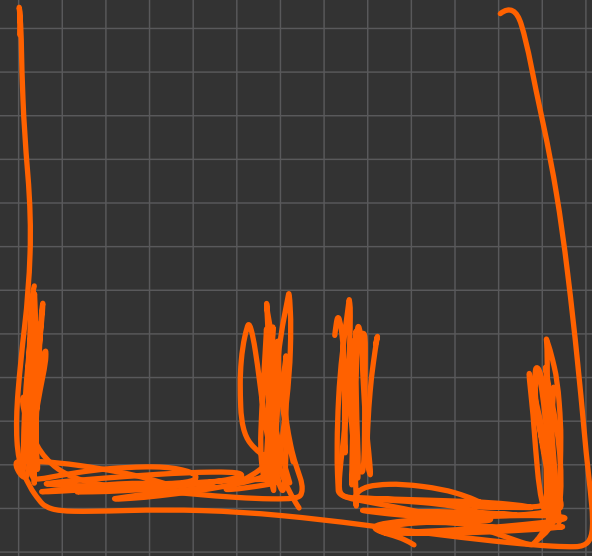
$dp(i, 1)$   $\swarrow$





1x2

$dp(0,0)$



$dp(0,1)$

F.S

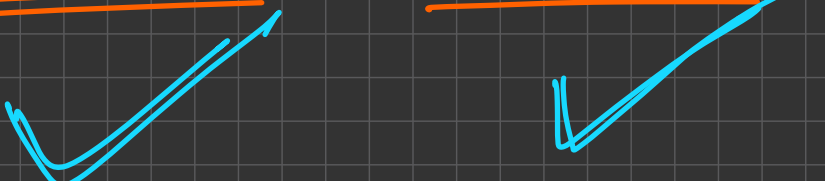
$dp(0,0) + dp(0,1)$

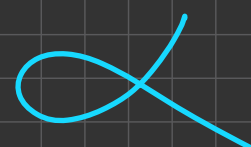


$11n^2$

# Space Optimization

- What other state does our current state depend on?
- Do we need answers to all subproblems at all times?
- Well, let's store only relevant states then!
- But wait, does this always work?
  - What if the final subproblem requires all the states?
  - What if we need to backtrack? [more on this in later lectures]

$$\underline{\underline{fib(n)}} = \underline{\underline{fib(n-1)}} + \underline{\underline{fib(n-2)}}$$


$$\underline{\underline{fib(n-3)}}$$


$$O(\underline{n^2})$$

State

$$n \leq \underline{\underline{10000}}$$

$$\Delta \quad O(n)$$

$$\underline{\underline{10^8}} =$$

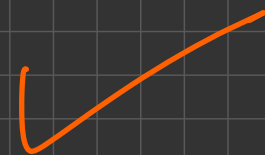
$$\underline{\underline{10^7}} \quad \underline{\underline{space}}$$

Tr

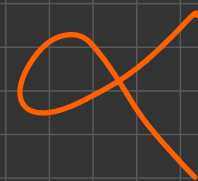
256 MB }  
512 MB }

256 MB

$O(10^8)$  time



$O(10^8)$  space

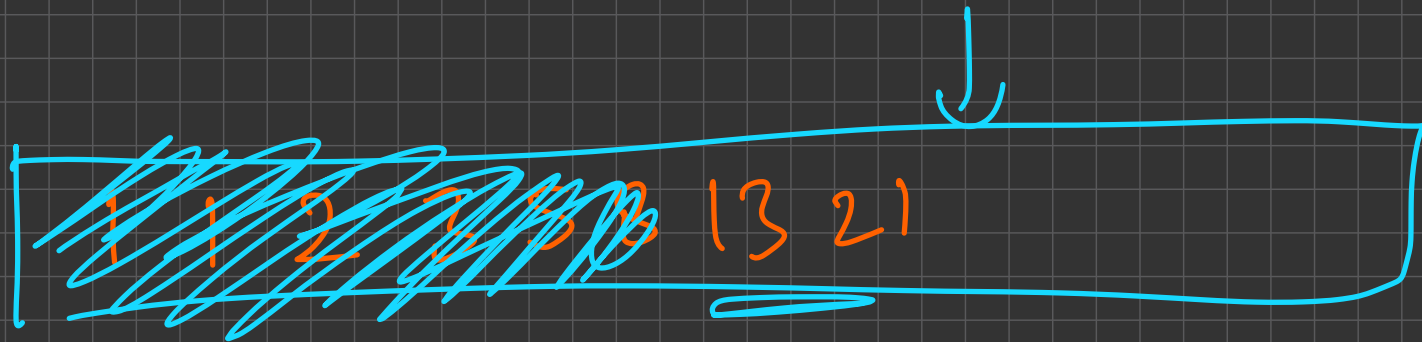


$O(n^2)$  state &  $O(1)$  T-T

OR

$n \leq 10^4$

$O(n)$  state &  $O(n)$  T-T

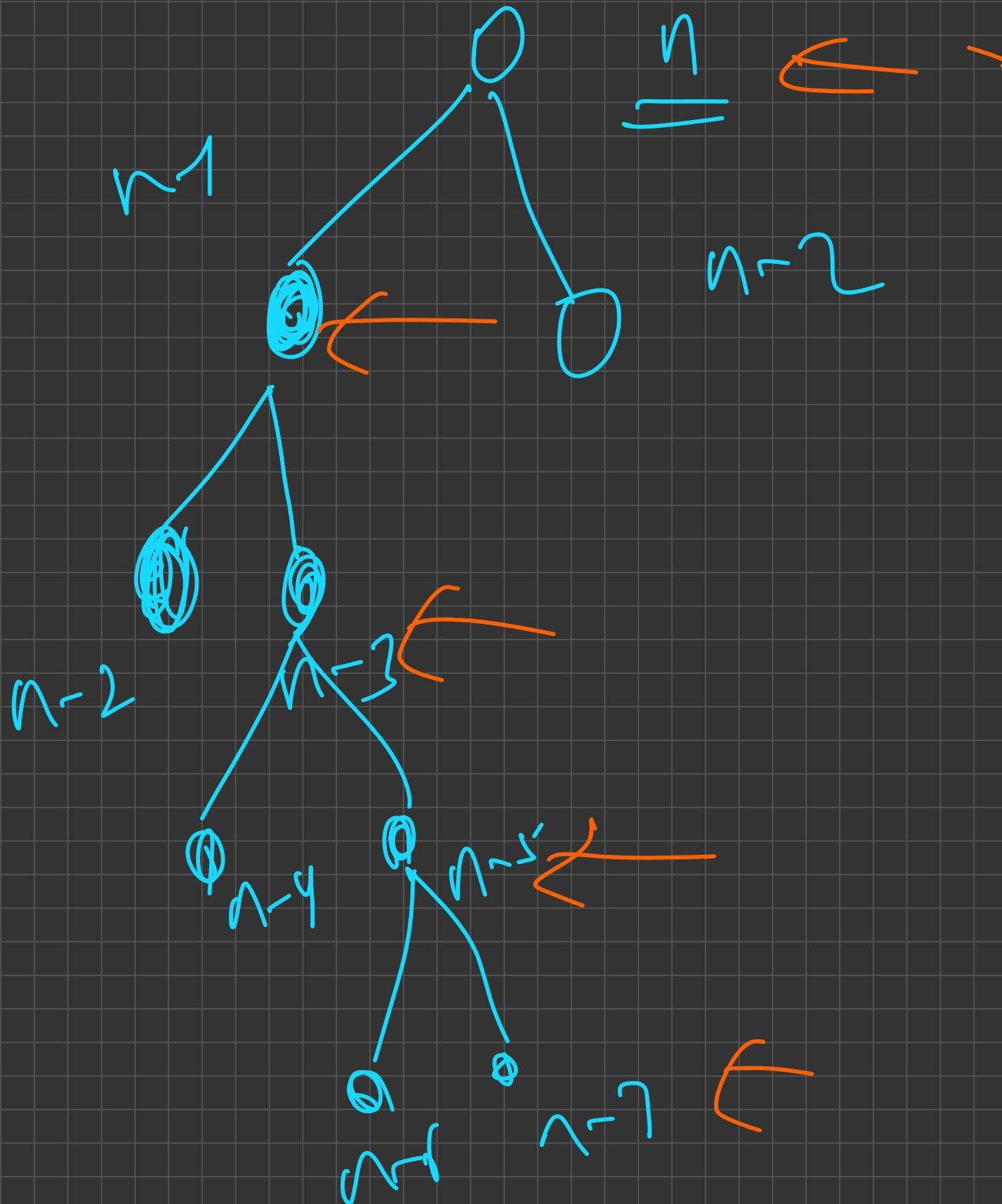


prev - prev

prev

$O(n)$

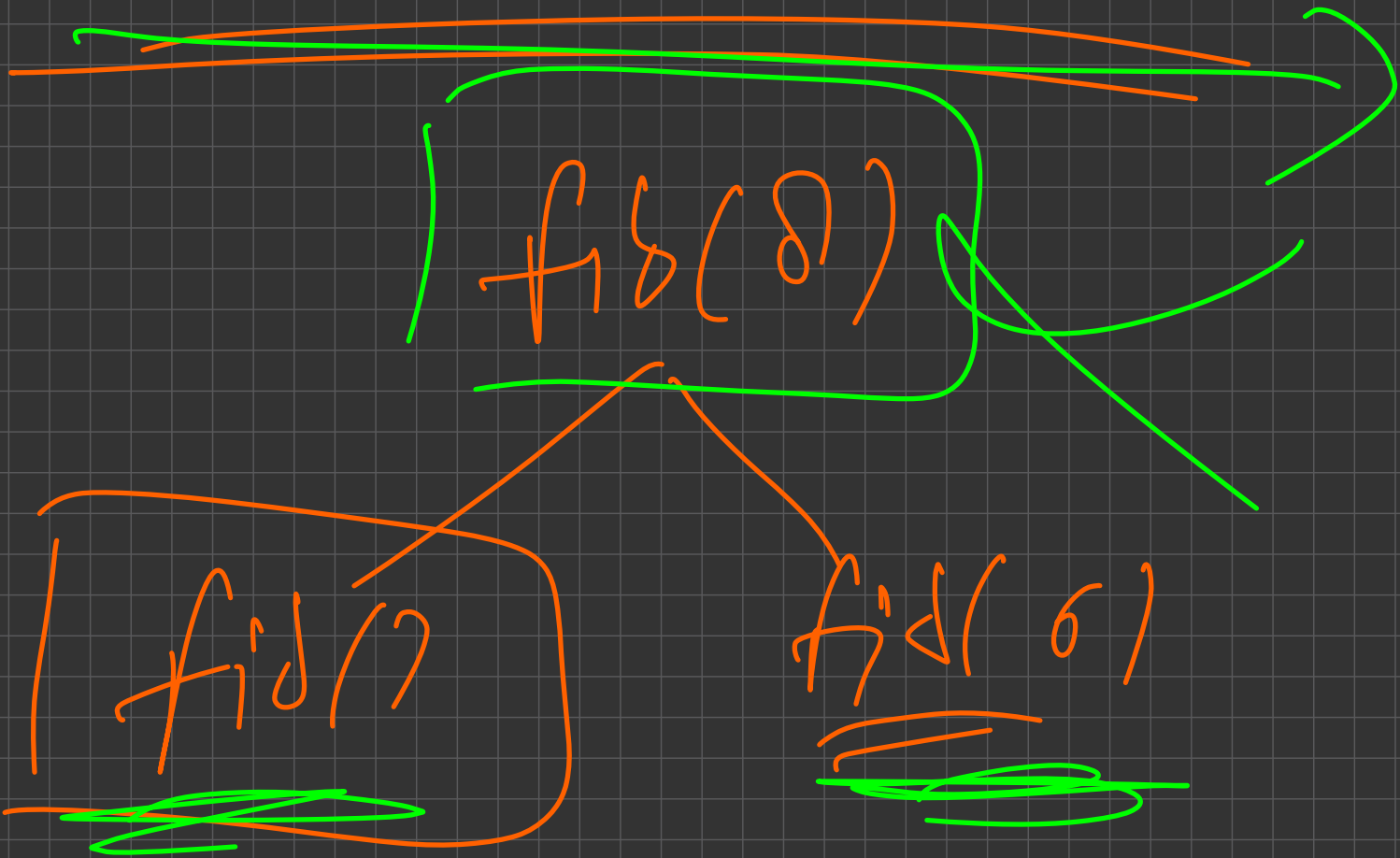
$$\text{current} = \underline{\underline{\text{prev}}} + \underline{\underline{\text{prev} - \text{prev}}}$$







0 → 1 → 2 → 3



$a = 1 \rightarrow f_{11}(1)$

$s = 1 \rightarrow f_{11}(2)$

$c = ? \rightarrow f_{11}(3)$

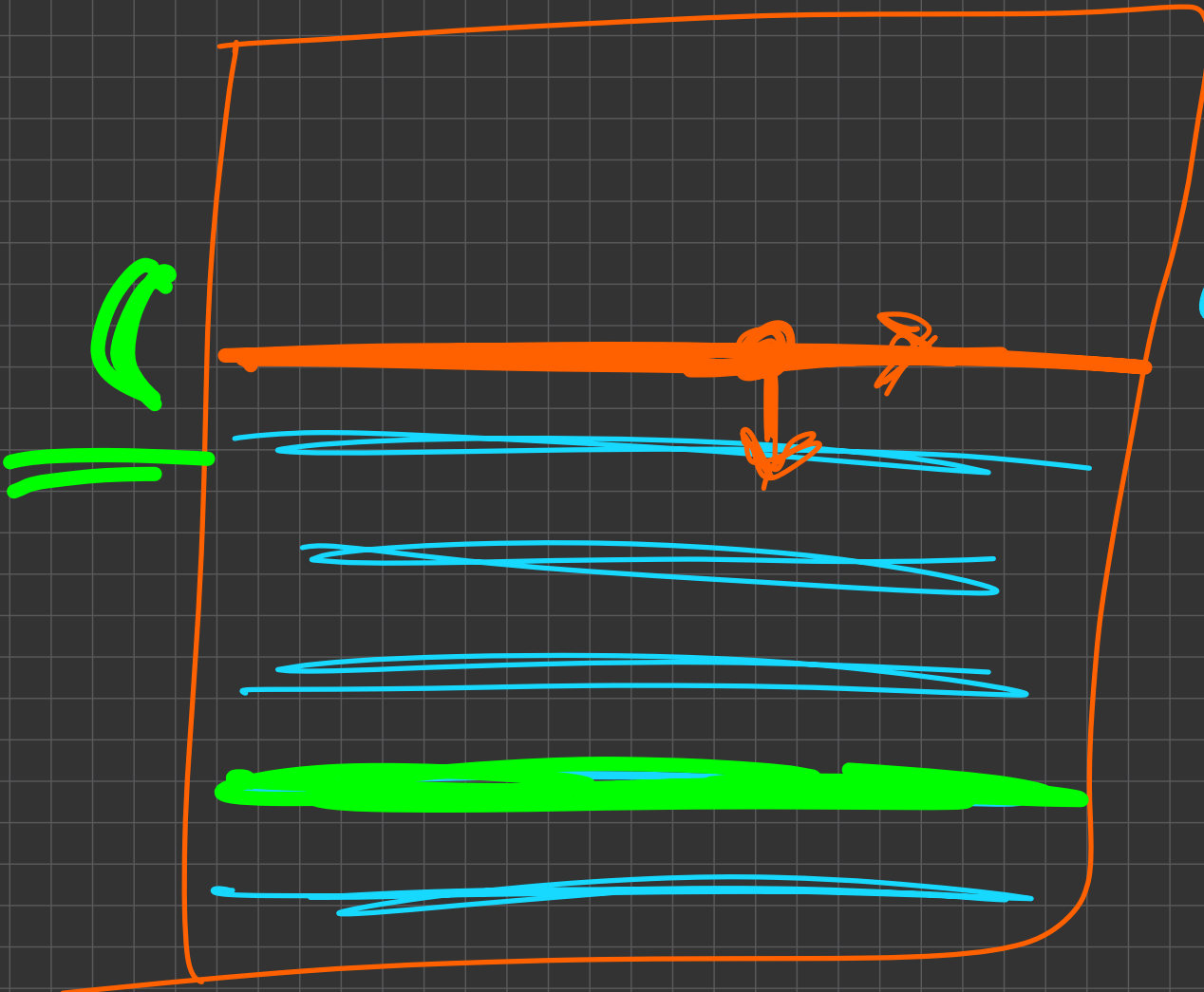
for (int i = 1; i ≤ n; i++)

$c = a + s$

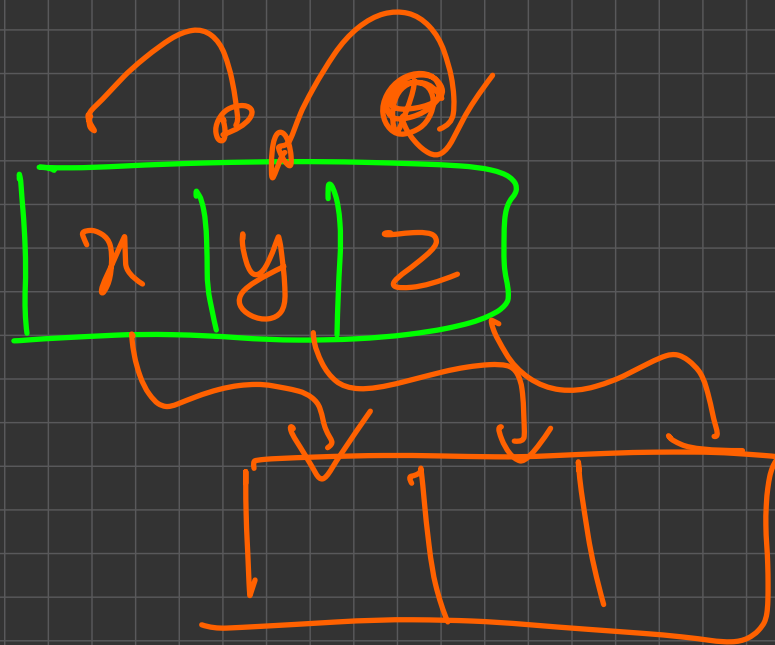
$a = s$

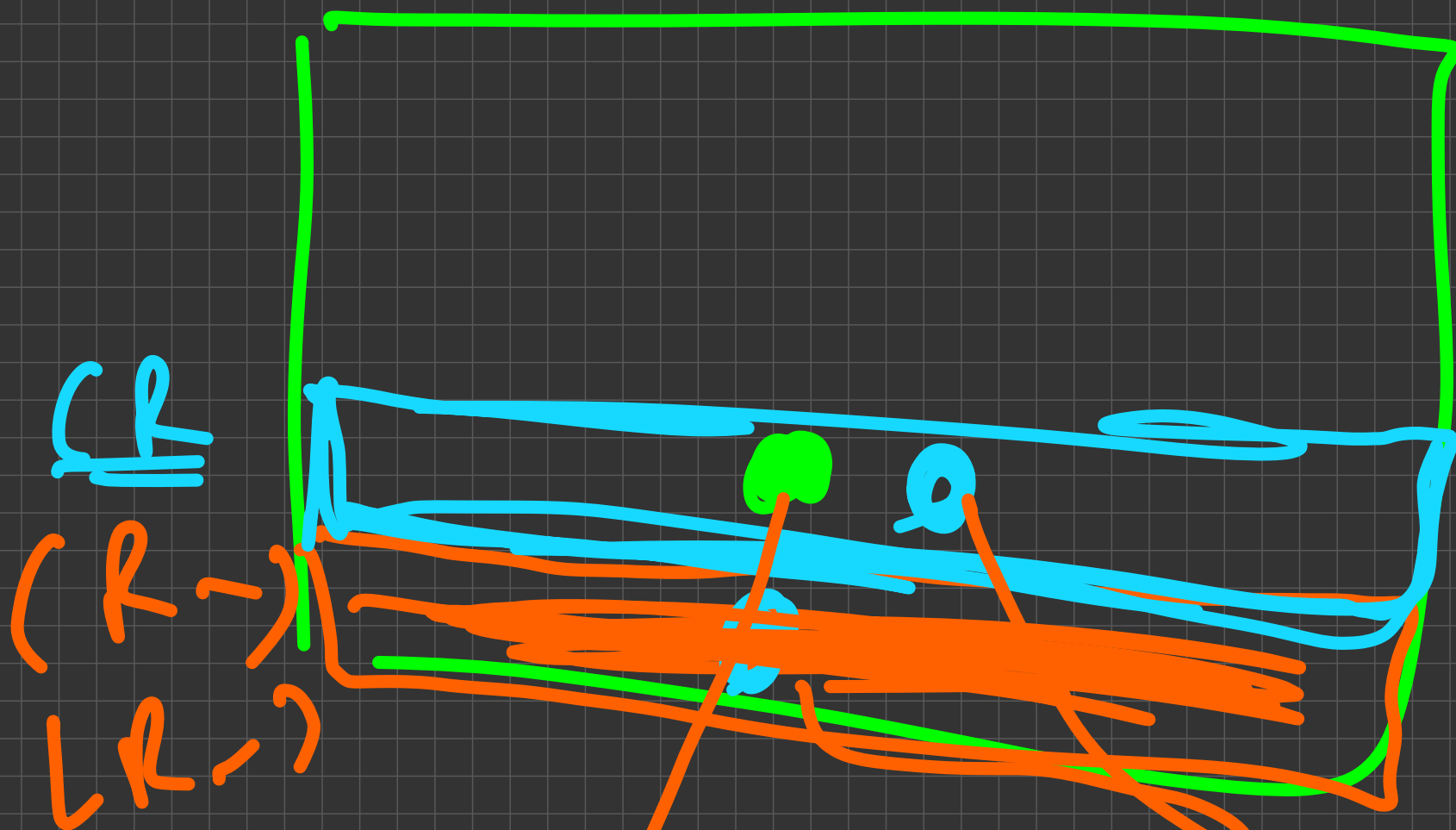
$s = c$

c

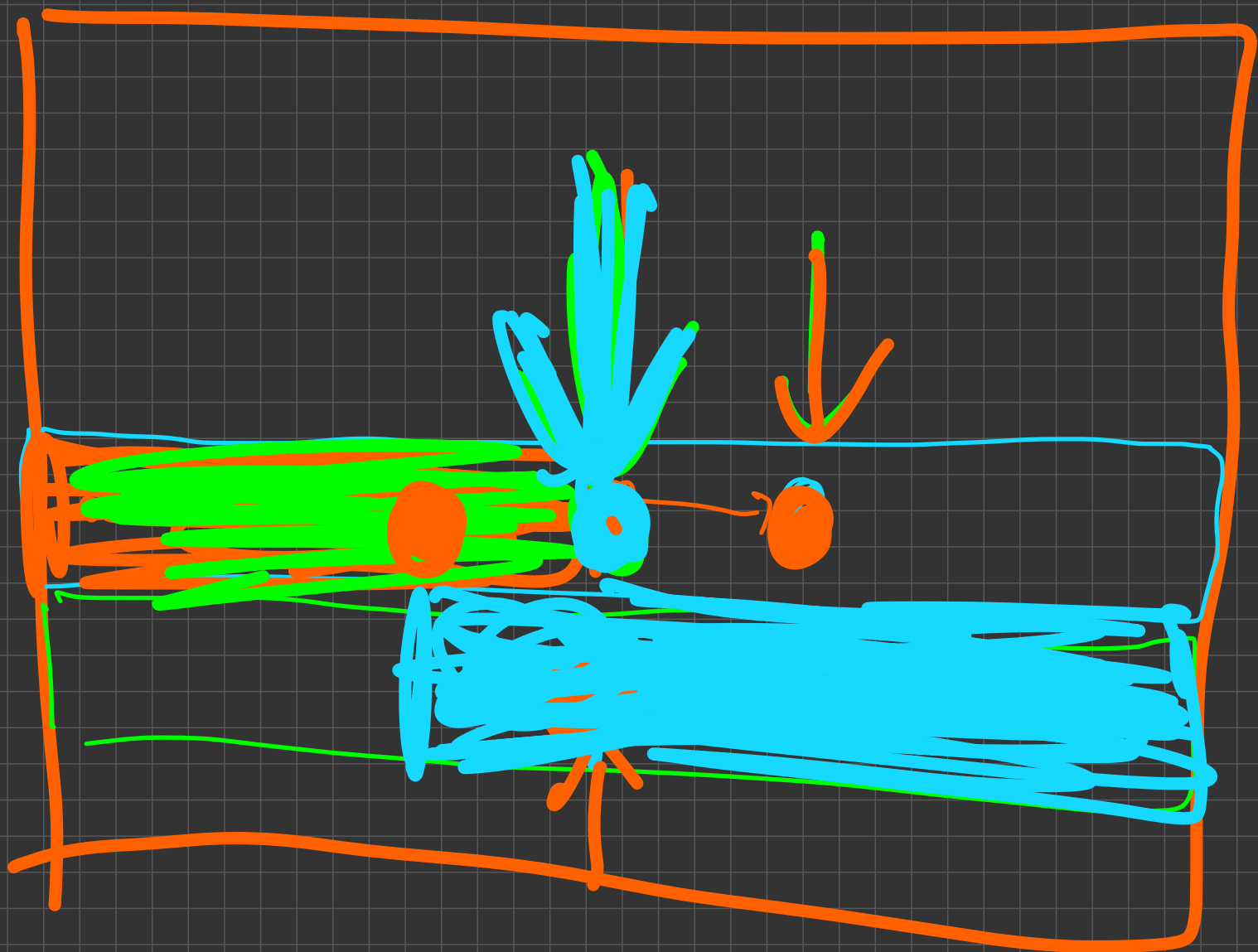


$dp(i, j) =$   
min sum path  
from  $(i, j)$   
to  $(n-1, m-1)$





$$CR(c) = \min \{ \dots \}$$



- Fibonacci Problem

- $dp[i]$  depends on  $dp[i - 1]$ ,  $dp[i - 2]$

- Grid Problem

- $dp[i][j]$  depends on  $dp[i + 1][j]$ ,  $dp[i][j + 1]$



# State Optimization

- Ask yourself do you need all the parameters in the dp state?
- If you have  $dp[a][b][c]$ , and  $a + b = c$ , do you need to store  $c$  as a parameter or can you just compute it on spot?
- If you can compute a parameter in dp state from other parameters, no need to store it.
- Which parameters should you remove? Highest

# Transition Optimization

- Observe the transition equation.
- Can you do some pre-computation to evaluate the equation faster?
- Using clever observations.
- Using range query data structures