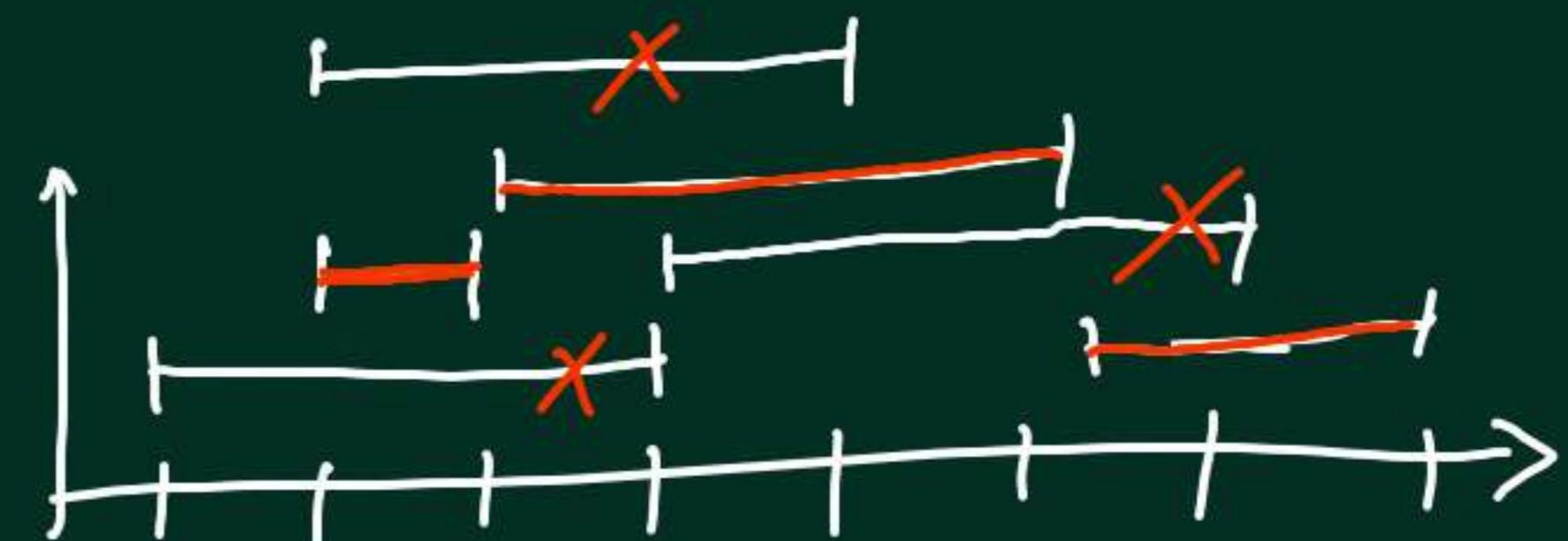


# Interval Scheduling

Given a set of intervals, find the largest subset of disjoint intervals.

Greedy approach: min finish time



$\mathcal{I}$   $\leftarrow$  input set

$I_0 \leftarrow \min \text{ finish time } \mathcal{I}$

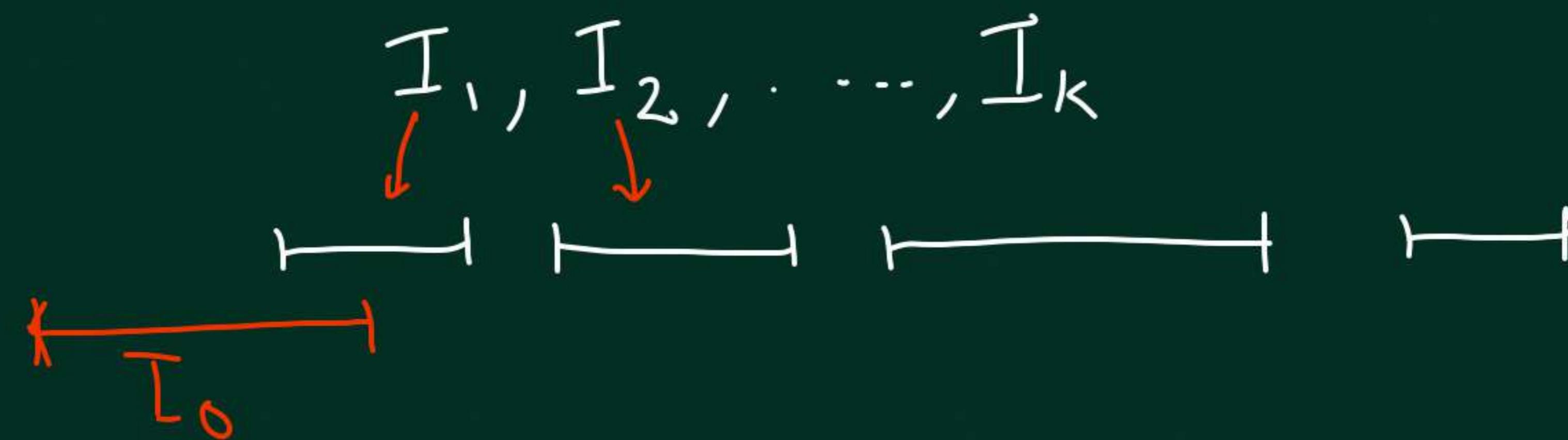
sort w.r.t.  
finish time

$\mathcal{I}' \leftarrow \mathcal{I} - \{ \text{intervals intersecting with } I_0 \}$

Output  $I_0 + \text{GreedyAlgo}(\mathcal{I}')$

Claim: There always exists an optimal solution  
that contains  $I_0$  (earliest finish time)

Proof: Consider some optimal solution



New solution  $I_0, I_1, I_2, \dots, I_k$   
This is a valid solution.  $K$  intervals.

Claim: Greedy algorithm gives an optimal solution.

Proof ( induction on number of intervals in the input)

Base case:  $n=1$ . Clearly optimal.

Induction Hypothesis: for any set of  $n-1$  intervals the greedy gives an optimal so

Induction step:

Greedy optimal for  $n$  intervals.

$\mathcal{I}$

$I_0 \leftarrow \min \text{ finish time}$

$|\mathcal{I}'| \leq n-1$ .

$\mathcal{I}' \leftarrow \mathcal{I} - \{ I_0 \}$  intersecting

Output  $I_0 + \boxed{\text{Greedy}(\mathcal{I}')}$

Claim:  $I_0 + \text{Opt}(\underline{\mathcal{I}}')$  is an optimal solution for  $\underline{\mathcal{I}}$

Proof: There is an optimal solution for  $\underline{\mathcal{I}}$  which contains  $I_0$ .

Let this be  $I_0, J_1, J_2, \dots, J_l$

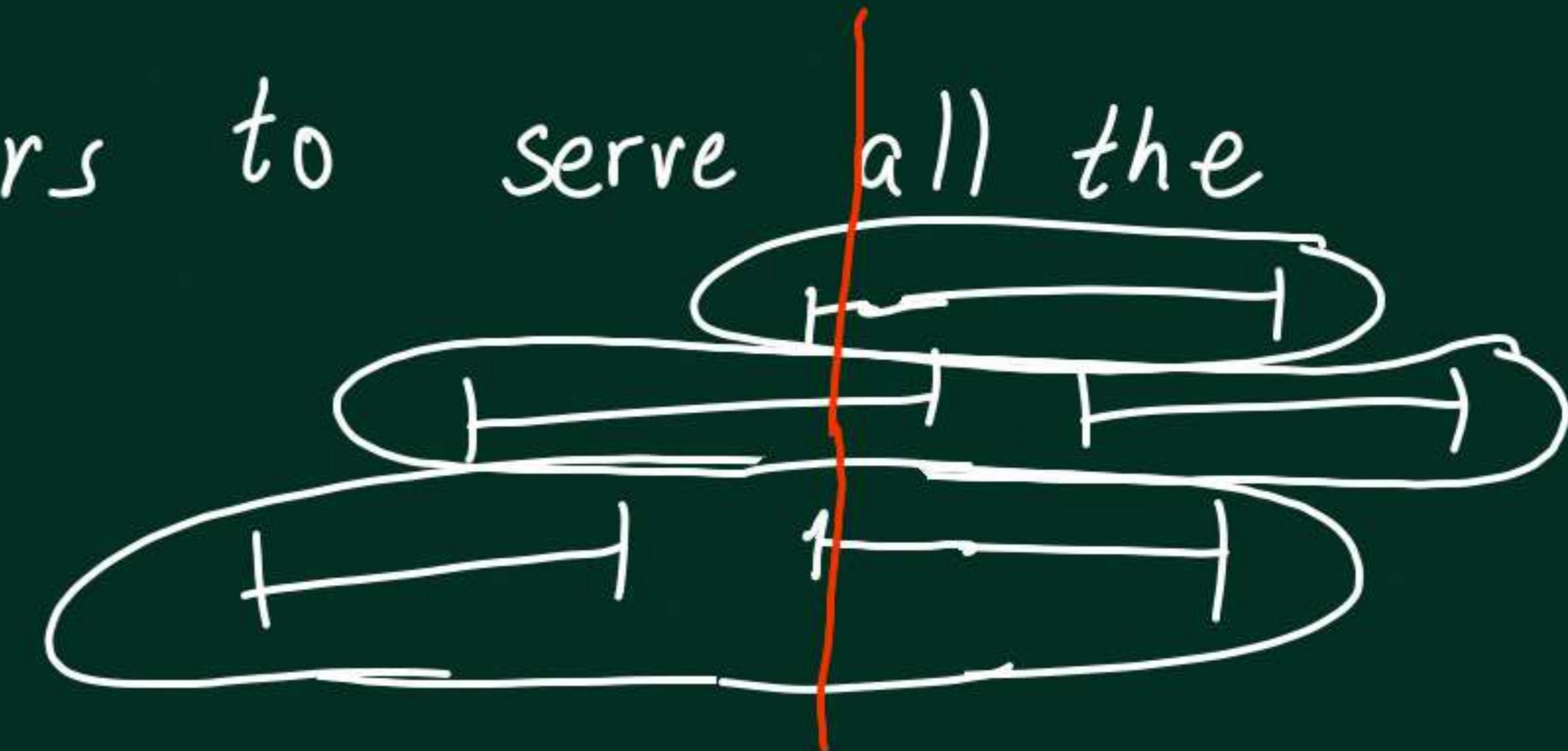
Obs:  $J_1, J_2, \dots, J_l$  disjoint from  $I_0$ .  
 $\in \underline{\mathcal{I}}'$

$J_1, J_2, \dots, J_l \leftarrow$  valid solution for  $\underline{\mathcal{I}}'$

$$l \leq |\text{Opt}(\underline{\mathcal{I}}')| \Rightarrow l+1 \leq |I_0 + \text{Opt}(\underline{\mathcal{I}}')|$$

## Problems

Minimum number of servers to serve all the requests (intervals)

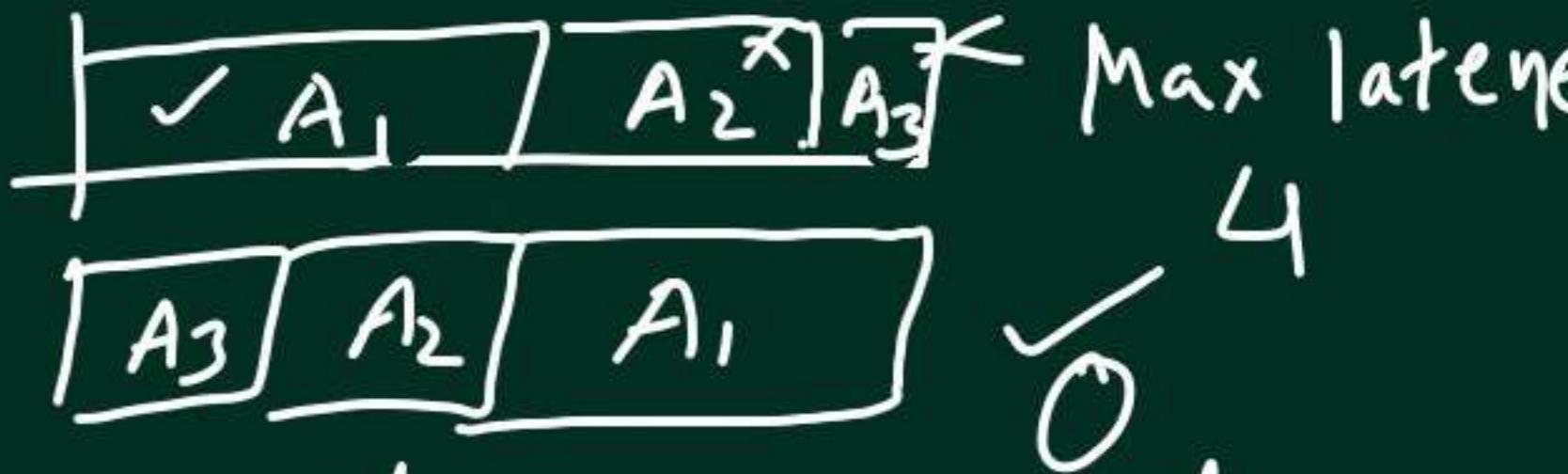


Minimum number of platforms

= Maximum no. trains at a given time instant.

Minimize max lateness

A<sub>1</sub> A<sub>2</sub> A<sub>3</sub>



n Assignments

deadlines: d<sub>1</sub>, d<sub>2</sub>, ..., d<sub>n</sub>

time : t<sub>1</sub>, t<sub>2</sub>, ..., t<sub>n</sub>

Que: Is it possible to do all assignments  
within deadlines.

Que: lateness

$$l_i = \max(f_i - d_i, 0)$$

	A <sub>1</sub>	A <sub>2</sub>	A <sub>3</sub>
time	3	2	1
deadline	6	4	2

Que: Maximize the number of assignments within deadline.

## Largest duration Interval scheduling

Given a set of intervals , find a set of disjoint intervals with maximum possible total length

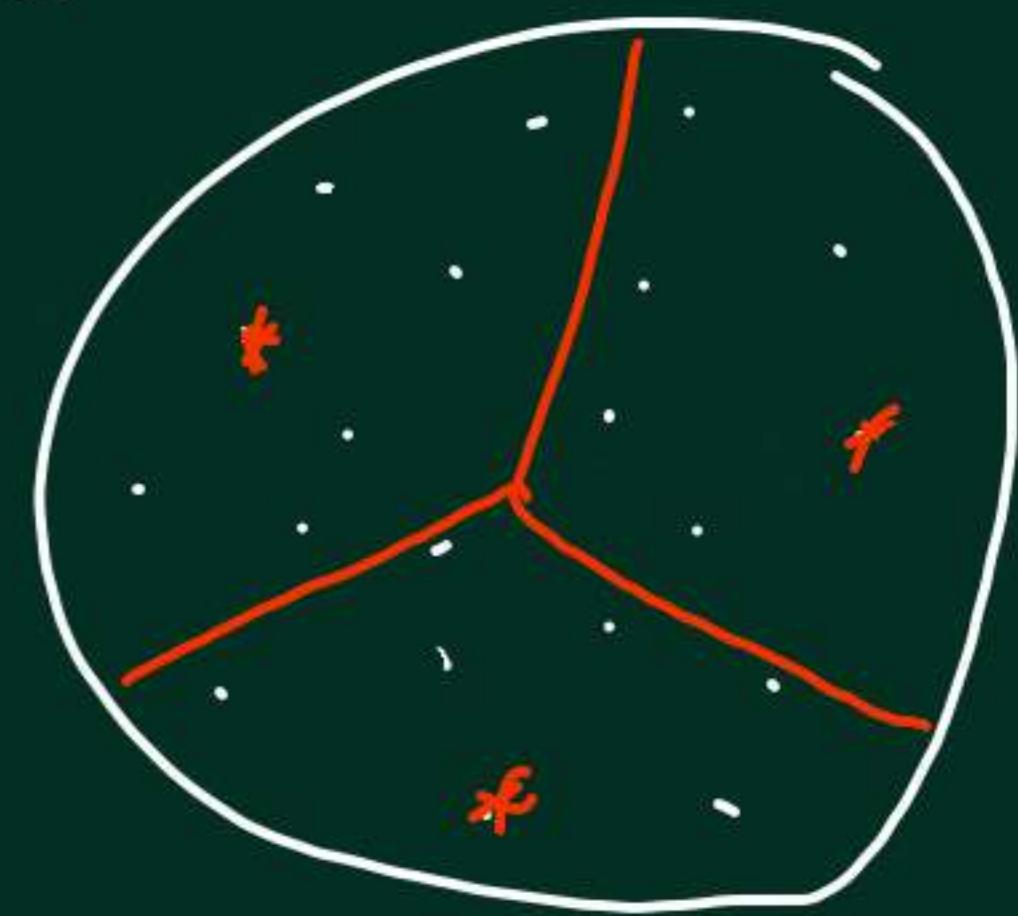
Greedy approaches:



- ① Max length first
- ② minimize the intersecting length.
- ③ minimize the number of intersections.
- ④ earliest start time
- ⑤ earliest finish time

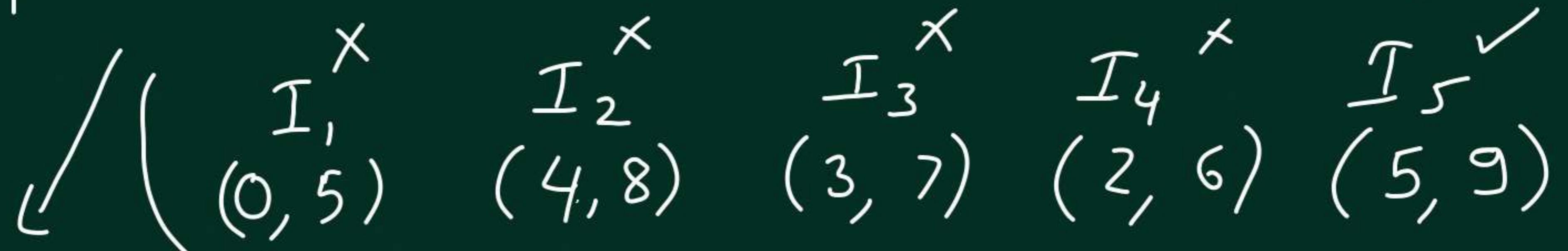
# Dynamic Programming

- Categorize all possible solutions into various categories
- Find an optimal solution from each category using the same algorithm recursively on some other input instances.
- Compare these optimal solutions and take the best.
- Store the solutions for all the inputs you have already solved.



# Longest duration interval scheduling

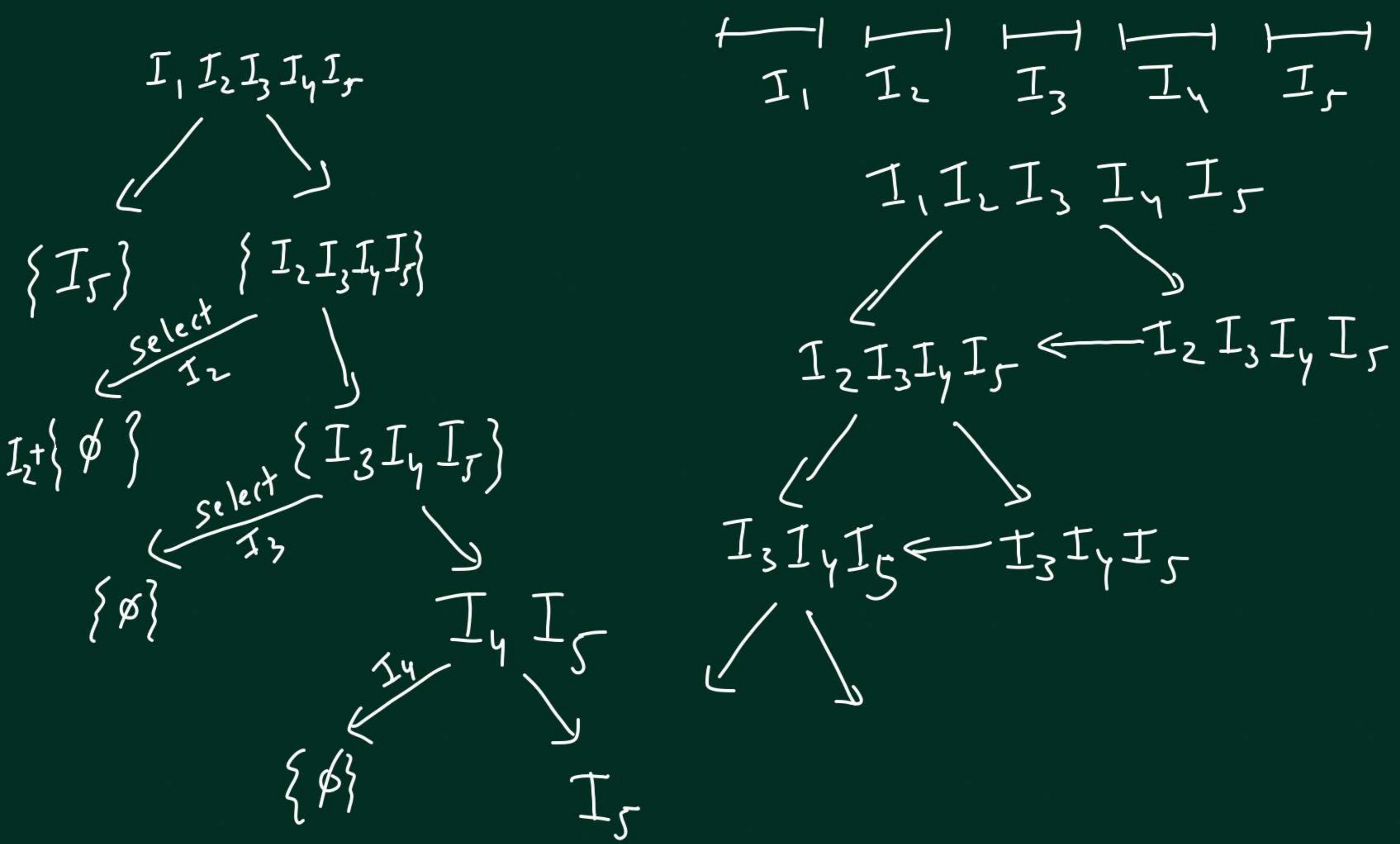
possible solutions  $\leftarrow$  subsets of disjoint intervals

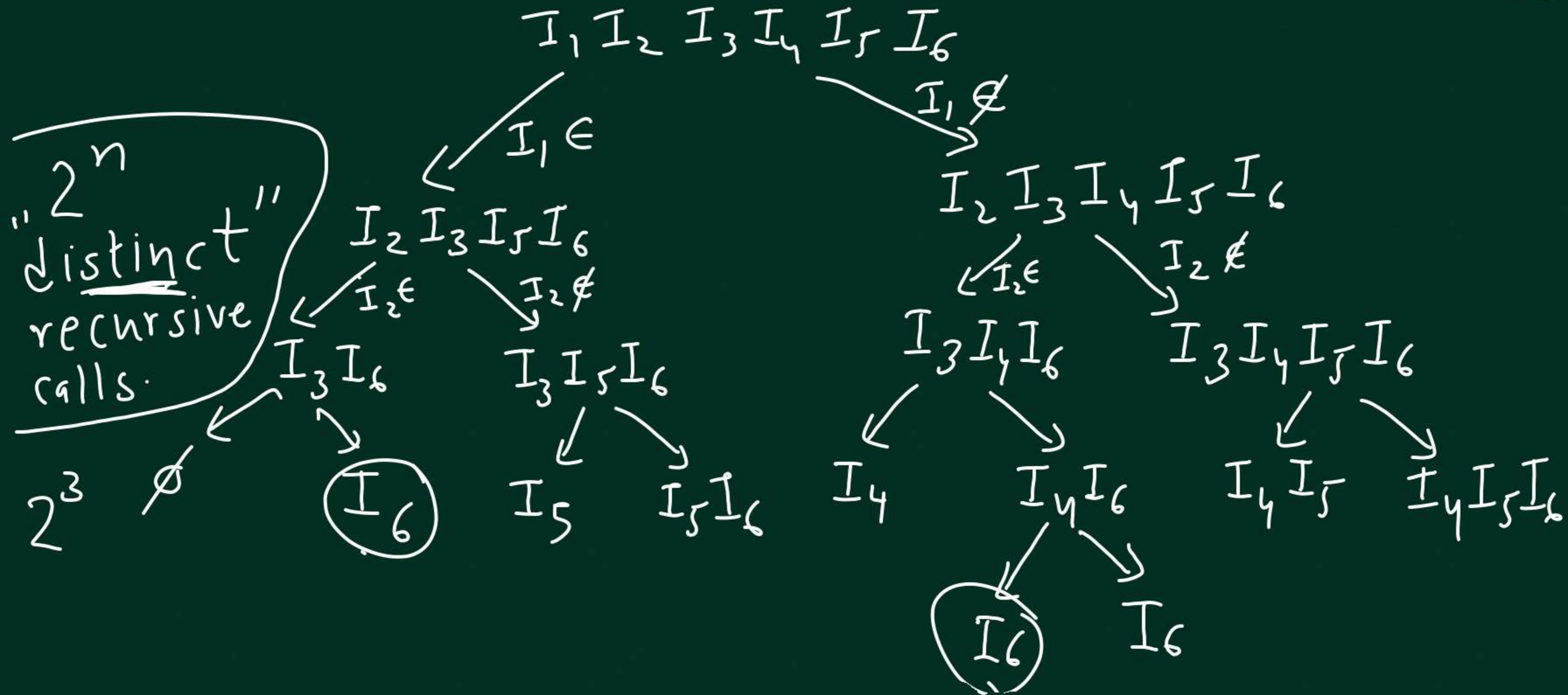
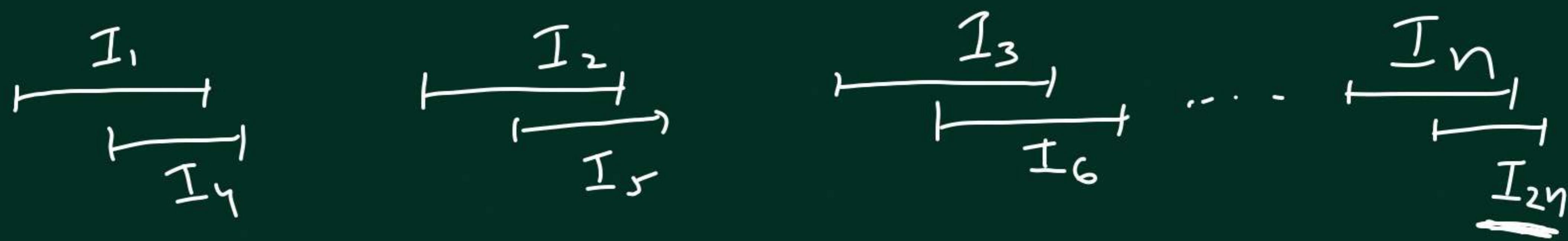


solutions  
which  
contain  $I_1$

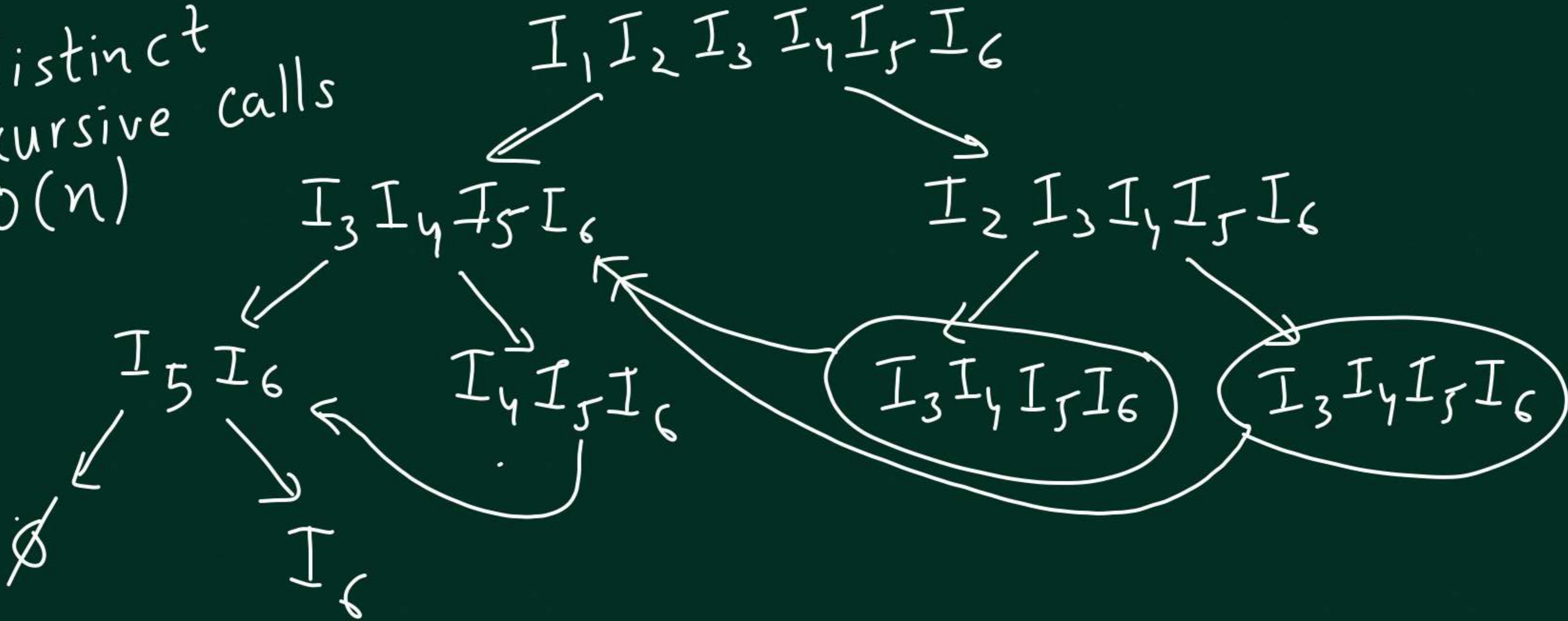
$ALG(I_5)$     $ALG(I_2, I_3, I_4, I_5)$

Best

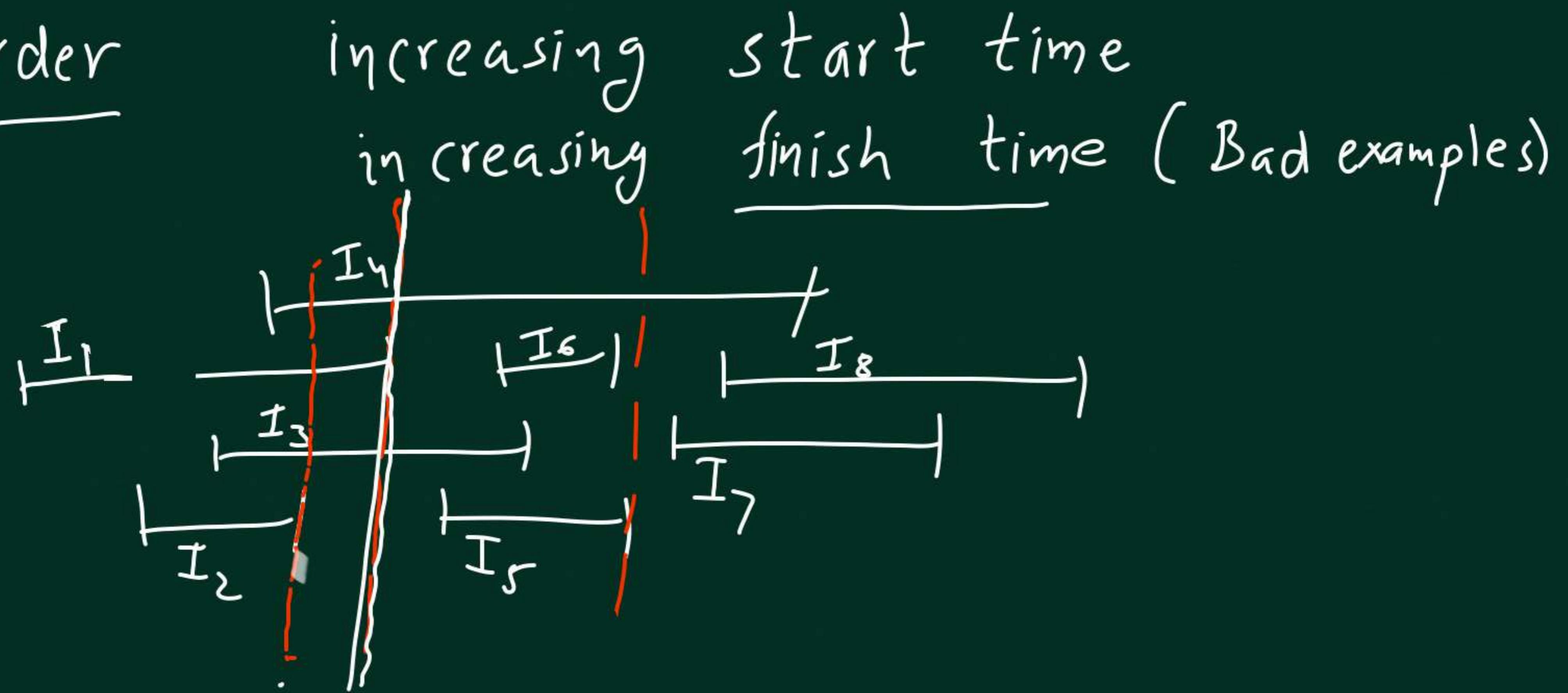




No. of  
"distinct"  
recursive calls  
 $O(n)$



Order



At most  $n$  distinct recursive calls.

$I_1 \dots I_n$



$j \leftarrow \text{first index}$

s.t.

$\underline{\text{start}(I_j)} > \underline{\text{finish}(I_1)}$

$\text{first}[k] \leftarrow \text{first index}$

j s.t.

$\underline{\text{start}(I_j)}$

$\underline{\text{finish}(I_k)}$

$\text{Opt}[j]$

Optimal solution

$(I_j, I_{j+1} \dots I_n)$

$\text{Opt}[i]$

For any k,

$\{I_k \dots I_n\}$

$\text{Opt}[k]$

$\text{Opt}[k+1]$

Max

$|I_k| + \text{Opt}[\text{first}[k]]$

## Interval Scheduling

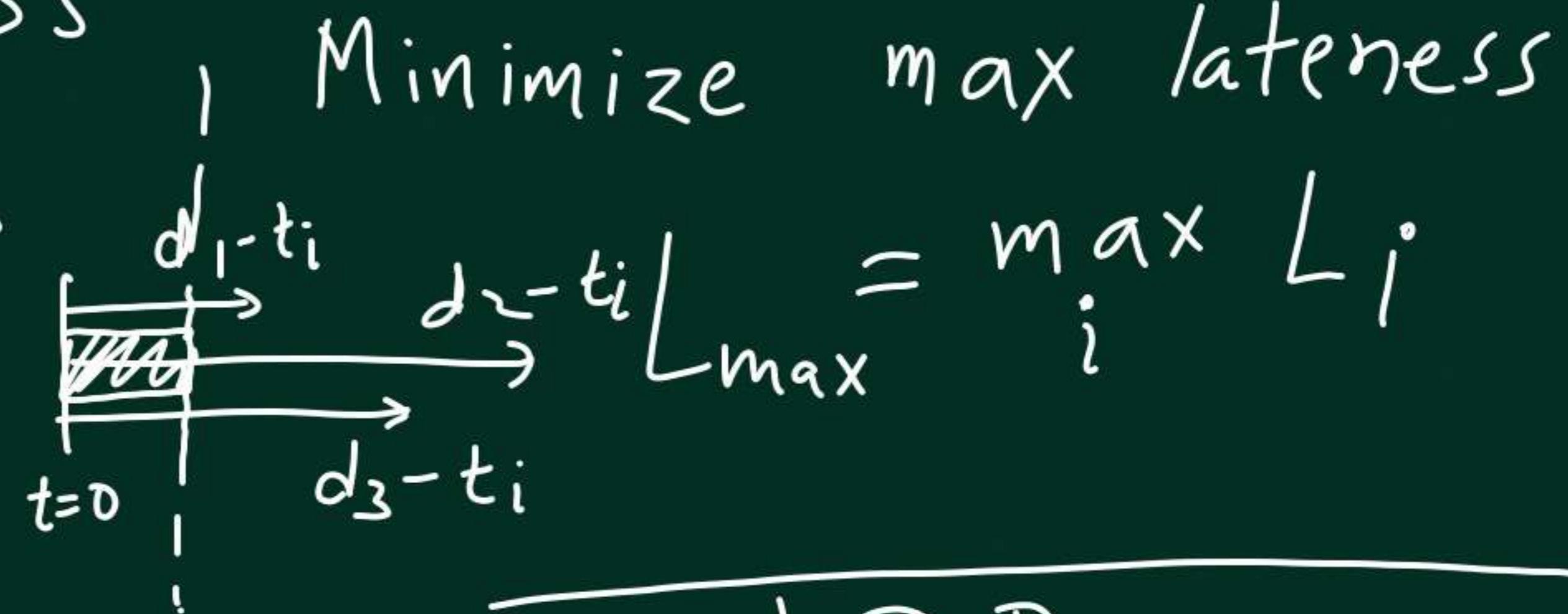
- ① Max no. of intervals (Greedy)
- ② Longest total duration (DP)  
 $O(n \log n)$
- ③ Max total weight of selected Intervals (DP)
- ④ Count the number of disjoint subsets of intervals (DP)

# Maximum Lateness

n assignments

deadlines  $d_1, \dots, d_n$

time  $t_1, t_2, \dots, t_n$



Lateness

of Assign i

$$L_i = \max \{0, f_i - d_i\}$$

↑

finish  $A_i$

DP

possible solutions is

$$\frac{n!}{A_1 \cdot A_2 \cdot \dots \cdot A_n}$$



$A_1$  is first

$A_2$

first

$A_n$  is first

Suppose  $A_i$  is scheduled at the first position.

$$\text{Opt}(\{A_1, A_2 \dots, A_n\})$$
$$\min_i \left\{ \max \left\{ \max \left\{ t_i - d_i, 0 \right\}, \text{Opt} \left( \underbrace{\{A_1, A_2 \dots, A_{i-1}\}}_{d-t_i}, \underbrace{A_{i+1} \dots A_n}_{d-t_i} \right) \right\} \right\}$$

No. of "distinct" recursive call

# Greedy

Schedule that assignment first which

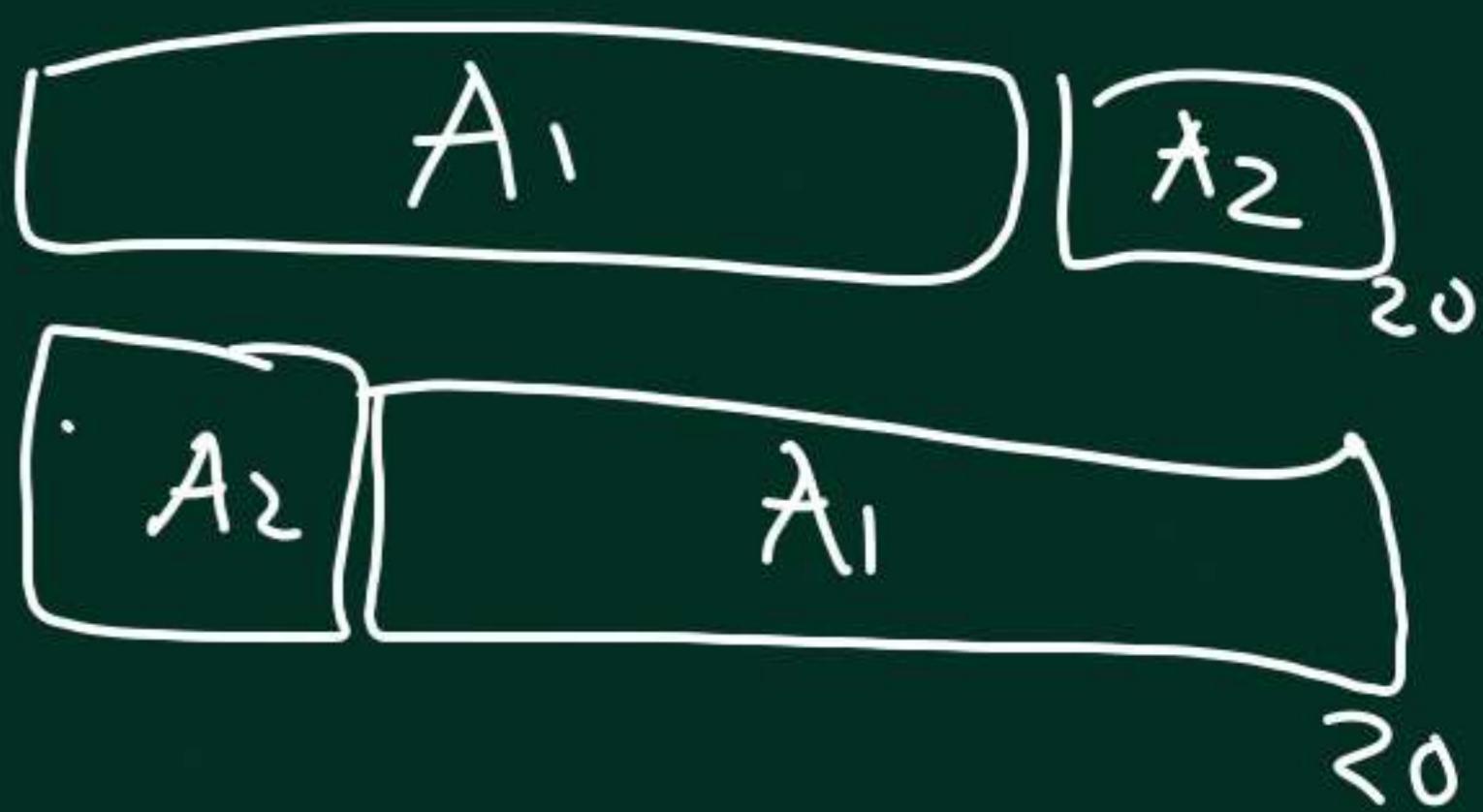
1 • minimizes  $d_i$

2 • ~~minimize  $t_i$~~

3 • ~~minimize  $d_i - t_i$~~   $L = 15$

$$L = 1$$

	$A_1$	$A_2$
$t$	18	2
$d$	19	5
$d - t$	1	3



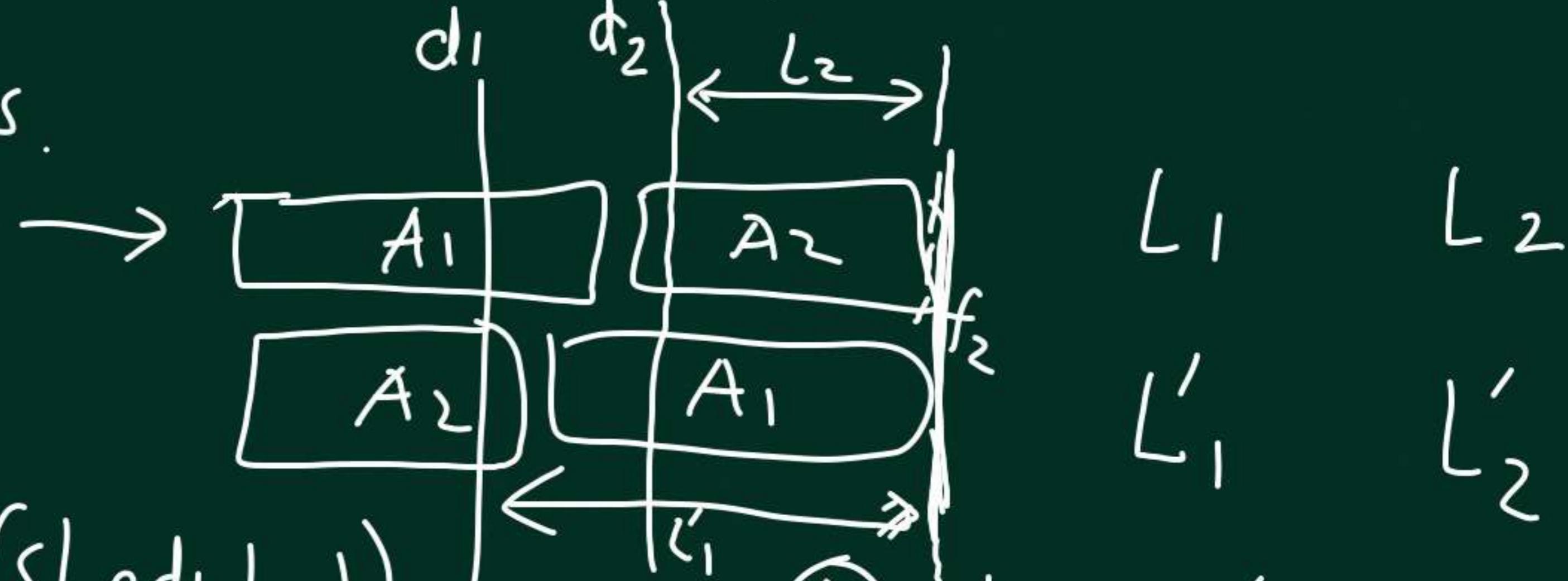
Schedule in the increasing order of deadlines.

Why is this optimal?

$$d_1 \leq d_2 \leq \dots \leq d_n$$

Two Assignments.

$$d_1 \leq d_2$$



Lateness (Schedule 1)

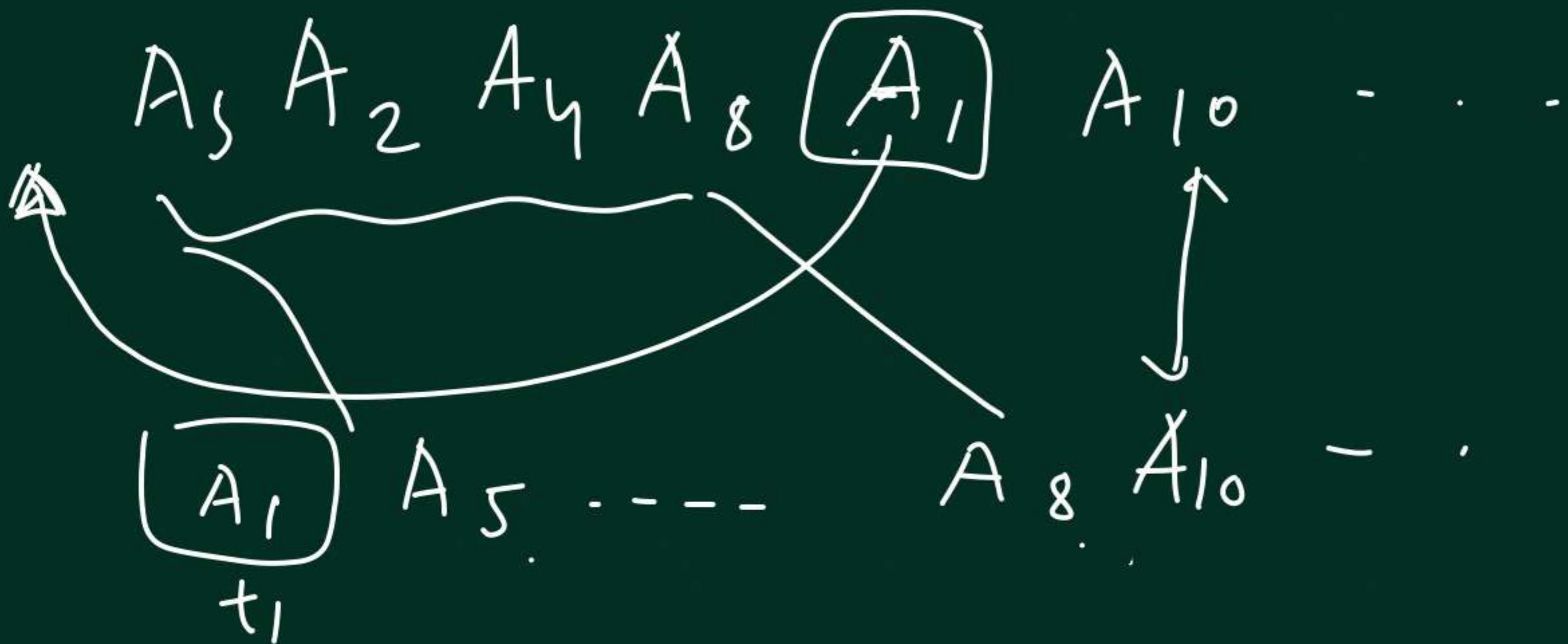
$\leq$  Lateness (Schedule 2)

①

$L_1 \leq L'_1$

②

$L_2 \leq L'_1$



$A_5 \ A_2 \ A_4 \ A_8 \ A_1 \ A_{10} \ \dots$   
 $\underline{A_5 \ A_2 \ A_4 \ A_1 \ A_8 \ A_{10}}$

lateness improves.

## Iterated Matrix Multiplication

$$\begin{bmatrix} 2 \times 3 \\ 2 \times 3 \end{bmatrix} \begin{bmatrix} 3 \times 4 \\ 3 \times 4 \end{bmatrix} \begin{bmatrix} 4 \times 5 \\ 4 \times 5 \end{bmatrix} = 2 \times 5$$

$$M_1 M_2 M_3 = (M_1 M_2) M_3 = M_1 (M_2 M_3)$$

$$\begin{array}{ccc} A & B & = C \\ p \times q & q \times r & p \times r \\ \boxed{\quad} \boxed{1} \boxed{x} & & \end{array} \quad \left. \begin{array}{l} pr \times q \text{ multi} \\ pr (q-1) \text{ additions} \end{array} \right\} O(prq)$$

$$\begin{array}{ccc}
 M_1 & M_2 & M_3 \\
 2 \times 3 & 3 \times 4 & 4 \times 5 \\
 \underbrace{(M_1 M_2)}_{2 \times 4} \cdot M_3
 \end{array}$$

$$\begin{array}{c}
 \textcircled{24} + 40 \\
 = 64
 \end{array}$$

Optimal order?

$$\begin{array}{r}
 \begin{array}{ccc}
 10 \times 3 & 3 \times 4 & 4 \times 5 \\
 \underbrace{120 + 200}_{\textcircled{60 + 150}}
 \end{array} \\
 \begin{array}{c}
 3 \times 5 \\
 \underbrace{3 \times 4}_{\textcircled{30}} \quad 4 \times 5 \\
 M_1 (M_2 M_3)
 \end{array}
 \end{array}$$

$$30 + 60 = 90$$

$$\begin{array}{cccc}
 P_0 & P_1 & P_2 & P_3 \\
 \textcircled{P_1 < P_2}
 \end{array}$$

$$\begin{array}{l}
 \frac{P_0 P_1 P_2 + P_0 P_2 P_3}{P_1 P_2 P_3 + P_0 P_1 P_3} = P_0 P_2 (P_1 + P_3) \\
 \underline{P_1 P_2 P_3 + P_0 P_1 P_3} = P_1 P_3 (P_0 + P_2)
 \end{array}$$

$$M_1 \quad M_2 \quad M_3 \quad \dots \quad M_n = M$$

$$P_0 \times P_1 \quad P_1 \times P_2 \quad P_2 \times P_3 \quad \dots \quad P_{n-1} \times P_n = P_0 \times P_n$$

Input  $P_0, P_1, P_2, \dots, P_n$

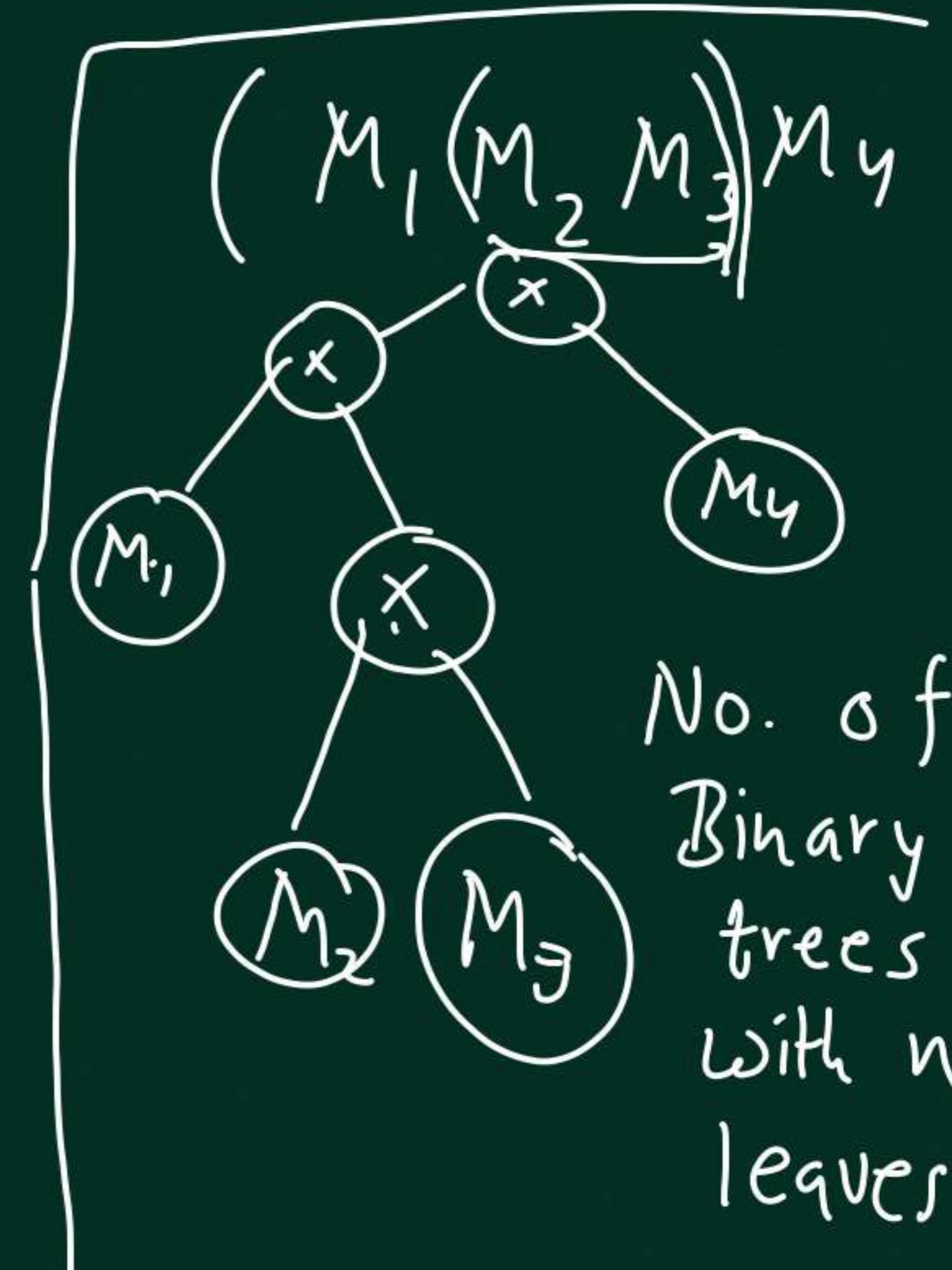
Output Best order

DP

categorizing solutions

$$M_1 (M_2, M_3) M_4$$

$$M_1 (M_2, M_3, M_4)$$



No. of  
Binary  
trees  
with n  
leaves.

$$M_1 M_2 (M_3 M_4) M_5 \dots M_n$$

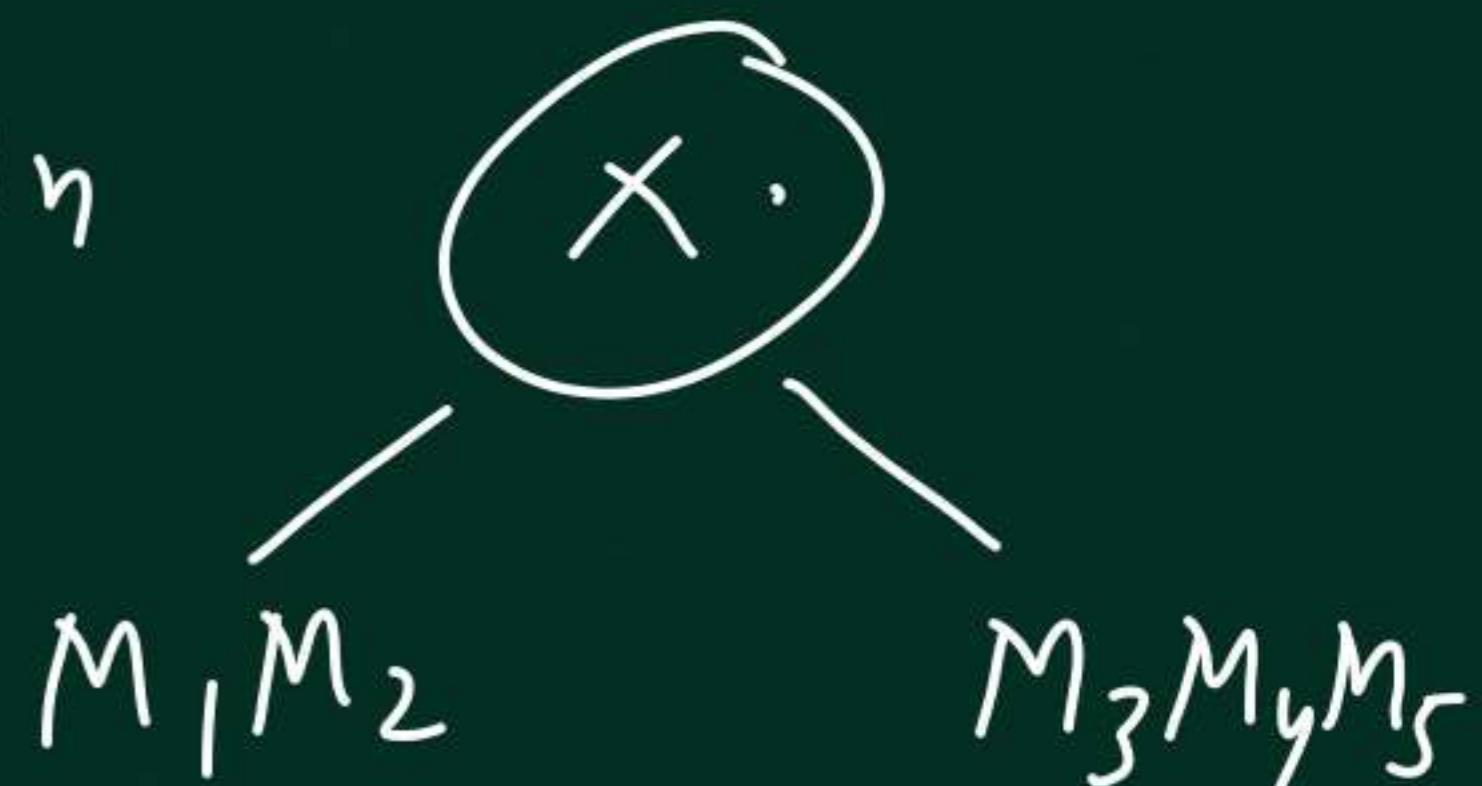
Classifying possible orders

~~①~~ first multiplication

② last multiplication

$$\xrightarrow{P_0 P_1 P_2 P_{i-2} P_{i-1} P_i P_{i+1} \dots P_n} M_1 M_2 \dots M_{i-1} (M_i M_{i+1}) \dots M_n \rightarrow M_1 M_2 \dots (M_i M_{i+1}) \dots M_n$$

$$\text{Opt}(P_0, P_1, \dots, P_n) = \min_i (P_{i-1} \cdot P_i \cdot P_{i+1} + \text{Opt}(P_0, P_1, \dots, P_{i-1}, P_{i+1}, \dots, P_n))$$



$$\text{Opt}(P_0, P_1, \dots, P_n)$$

$$= \min_i \left\{ P_{i-1} P_i P_{i+1} + \text{Opt}(P_0, P_1, \dots, P_{i-1}, P_{i+1}, \dots, P_n) \right\}$$

No. of distinct recursive calls?  $\ll \exp(n)$

$$\begin{aligned} & l_n \\ & \times (n-1) \\ & \times (n-2) \\ & \times (n-3) \end{aligned}$$

$$P_0, P_1, P_2, P_3, \dots, P_n$$

Every subsequence  
is possible



$$\frac{(M_1 M_2 \dots M_i) \times (M_{i+1} \dots M_n)}{P_0 P_1 \dots P_{i-1} P_i P_{i+1} \dots P_n}$$

Opt( $P_0, P_1, P_2, \dots, P_n$ )      No of Distinct recursive calls  $\leq \binom{n+1}{2}$

$$P_3 = \min_i \left[ P_0 P_i P_n + \text{Opt}(P_0, P_1, \dots, P_i) + \text{Opt}(P_i, P_{i+1}, \dots, P_n) \right]$$

Every recursive call  $\rightarrow P_k P_{k+1} \dots P_l$

$$\text{Opt}(P_K \dots - P_\ell)$$

$$= \min_i \left\{ P_K \cdot P_i \cdot P_{K-i} + \begin{cases} \text{Opt}(P_K - P_i) \\ \text{Opt}(P_i - P_\ell) \end{cases} \right\}$$

$O(n^3)$  time.

Implementation

Subset Sum problem.

$$\{ \boxed{6}, 2, -5, \textcircled{4}, \textcircled{-7}, 15, -20, -10, 8, 9 \}$$

Is there a subset whose sum is zero?

# Subset Sum problem. (Dynamic Programming)

$\{6, 2, \textcircled{-5}, 13, \textcircled{9}, \textcircled{-7}, 15, -20, -10, \textcircled{8}, 9\}$

Is there a subset whose sum is zero?

Trivial  $\leftarrow 2^n$

Subset  $\rightarrow$  will have an  
will not have an

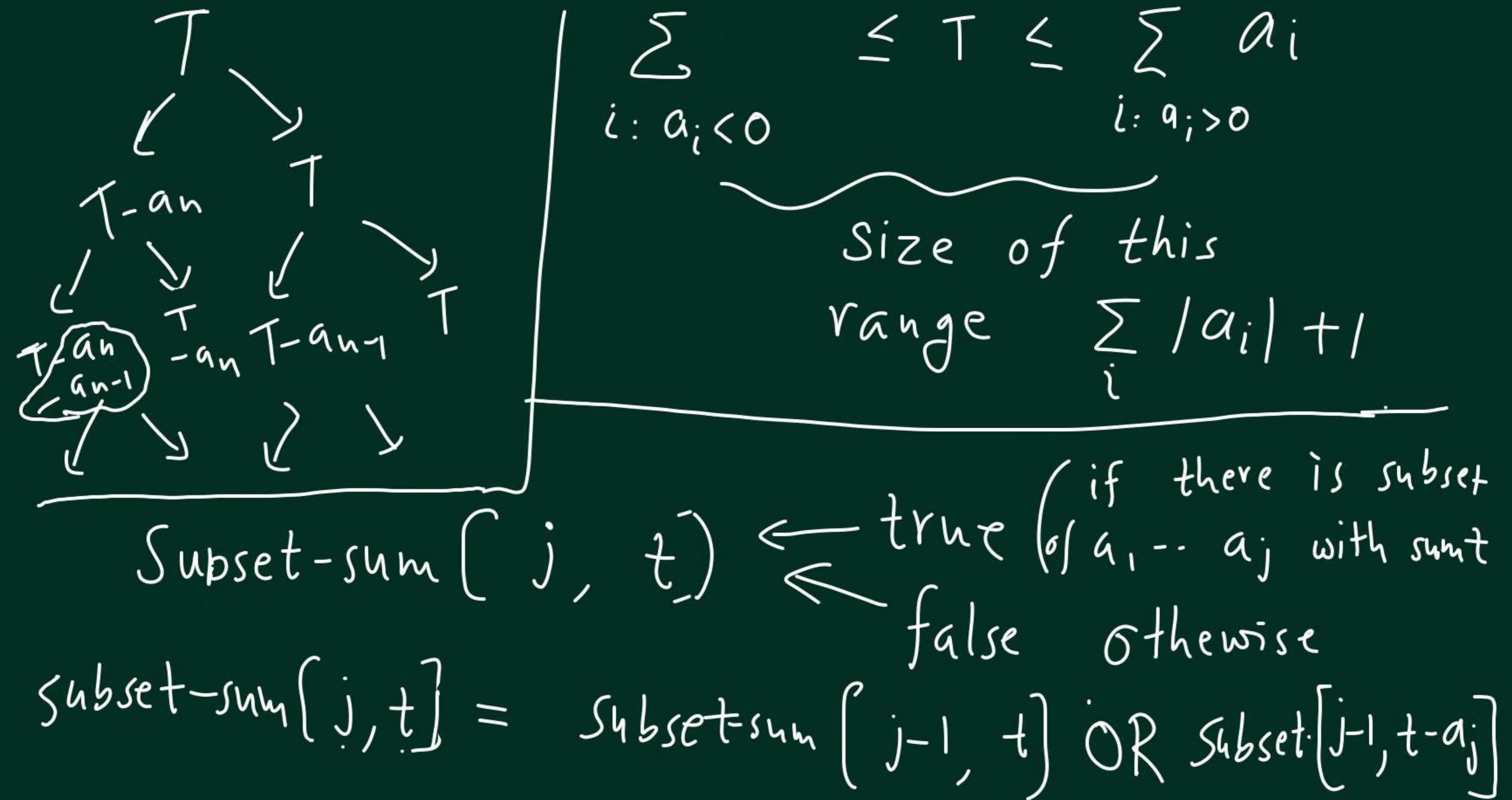
Is there a  
subset  
of  $a_1 \dots a_n$   
with sum  
 $a_n$

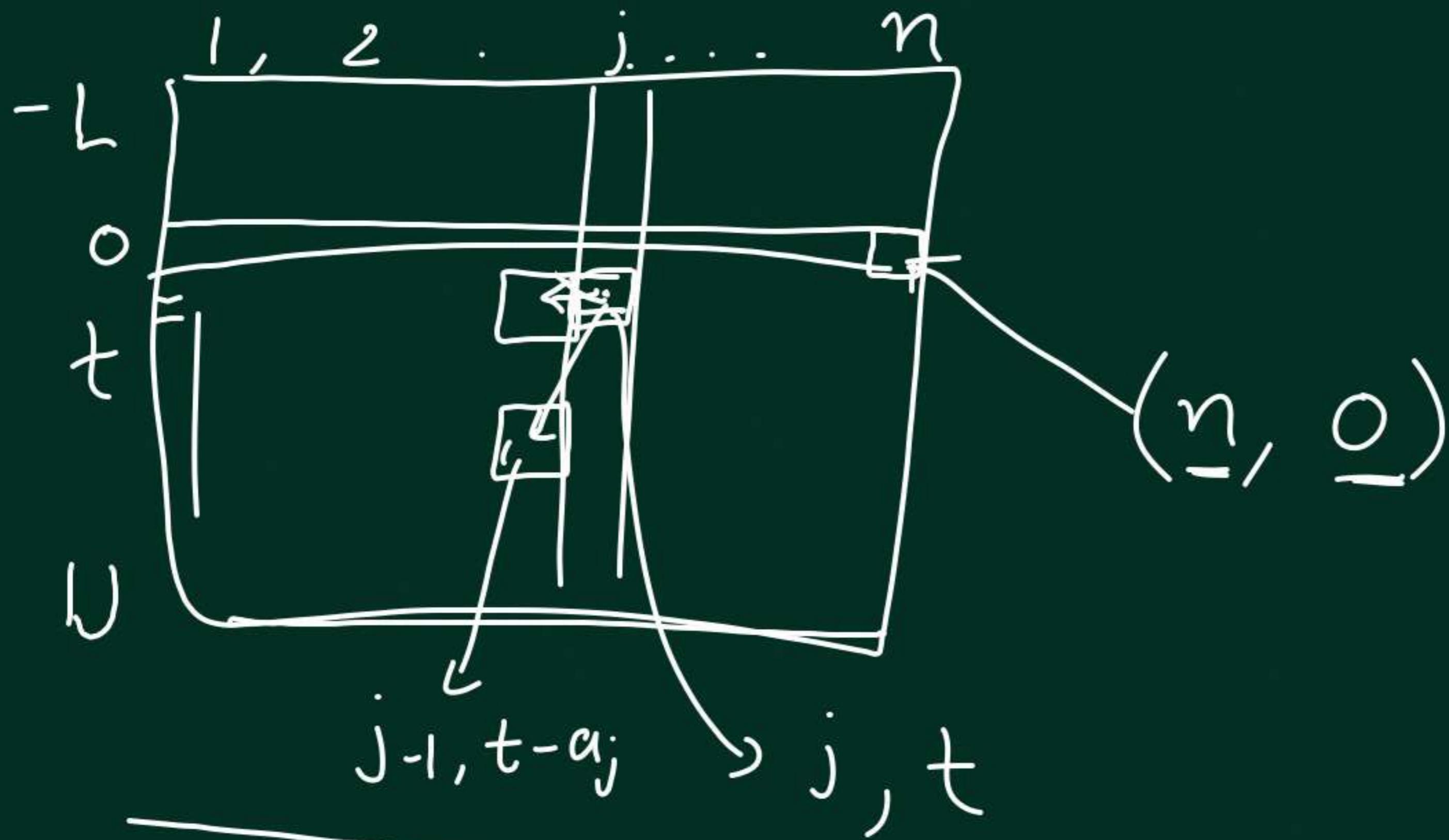
Is there a subset of  $a_1 \dots \underline{a_n}$   
with sum  $T$

Is there a subset  
of  $\underline{a_1 \dots a_{n-1}}$   
with sum  $T - a_n$

Is there a subset  
of  $a_1 \dots a_{n-1}$   
with sum  $T$ .

No. of distinct recursive calls?  
prefix  $\underline{n} \leftarrow (\underline{a_1, \dots, a_j}, \underline{\text{Target } t})$





Implementation

$n \downarrow$  numbers  
 input size  
 = total no. of bits

Complexity:  $O(n \times \sum |a_i|)$  =  $O(\underbrace{n \cdot n \cdot 2^d})$

If  $a_1, \dots, a_n \leftarrow l$  bit numbers  
 Input size  $n \cdot l$ . pseudo-polynomial time.

Subset-Sum  $\leftarrow$  NP-hard.

unlikely to have a polynomial time algorithm.

Greedy doesn't work

## Knapsack Problem

Value  $p_1, p_2, \dots, p_n$

Pseudopoly time.  
 $O(n \cdot W)$

Weights  $w_1, w_2, \dots, w_n$

Poly  $(n, \sum_i p_i)$

Pick the subset with max total value but total weight  $\leq W$ .

Fractional Knapsack.

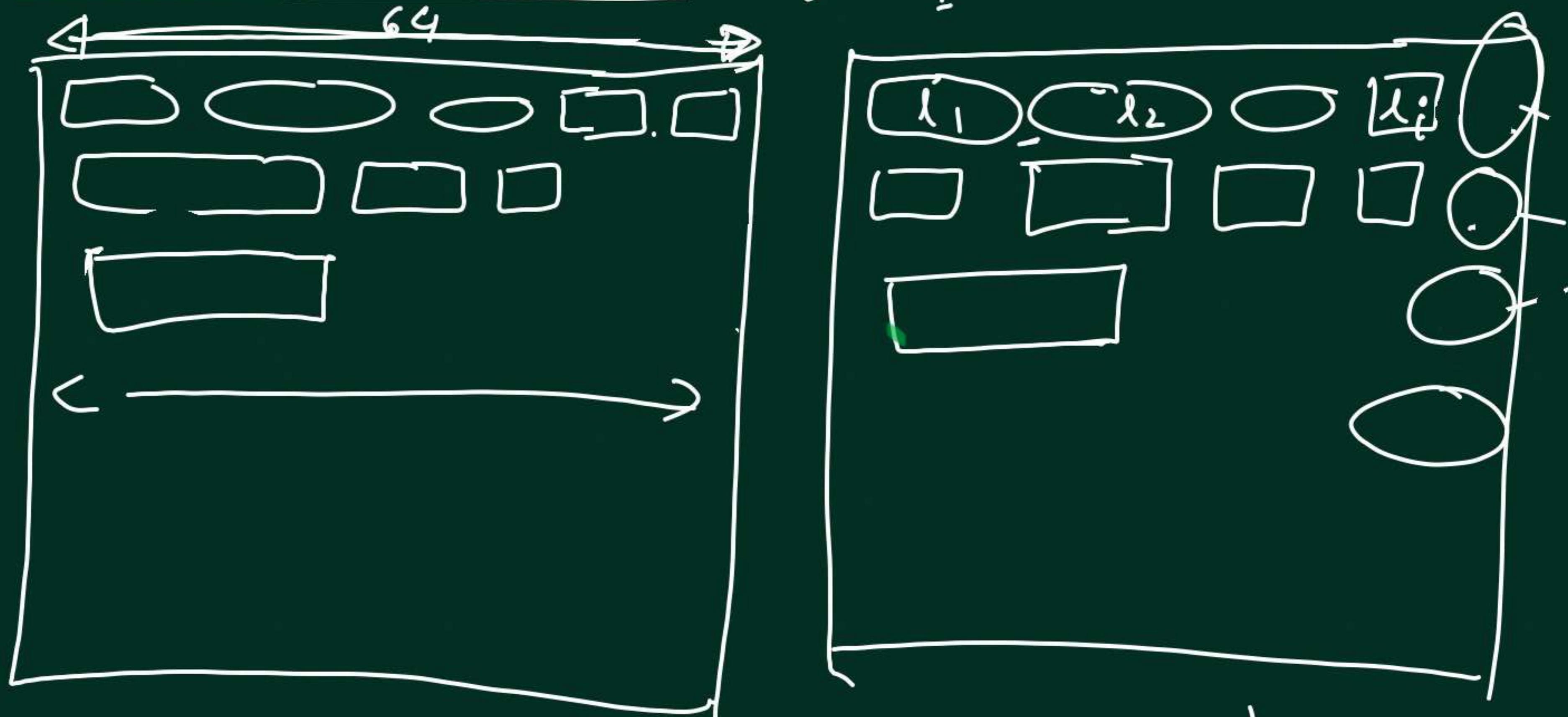
$$x_1, x_2, \dots, x_n \in [0, 1]$$

$$\max \sum_i p_i x_i$$

Subject to  $\sum_i w_i x_i \leq W$

Greedy algorithm works.

Balanced Margins.  $\text{Slack}_1 = L - l_1 - 1 - l_2 - 1 \dots - l_i$



Input  $\leftarrow$  sequence of word  
 $l_1, l_2, \dots, l_n$  } Limit per  
line  $L$ .

$L \leftarrow \text{linewidth}$

Input :  $l_1, l_2, \dots, l_n$

Line  $k$        $i^{\text{th}}$  to  $j^{\text{th}}$  word

$$\text{slack}_k = L - (l_i + 1 + l_{i+1} + 1 + \dots + l_{j-1} + 1 + l_j)$$

roughly balanced

$$\sum_k \text{slack}_k = \text{const.} \quad | \quad \text{Minimize} \quad \sum_{k=1}^q (\text{slack}_k)^2$$

find  $a, b, c \in \mathbb{Z}$

$$a + b + c = 14$$

minimize  $\underline{a^2 + b^2 + c^2}$

$$14 = 10 + 2 + 2$$

$$14 = 4 + 5 + 5$$

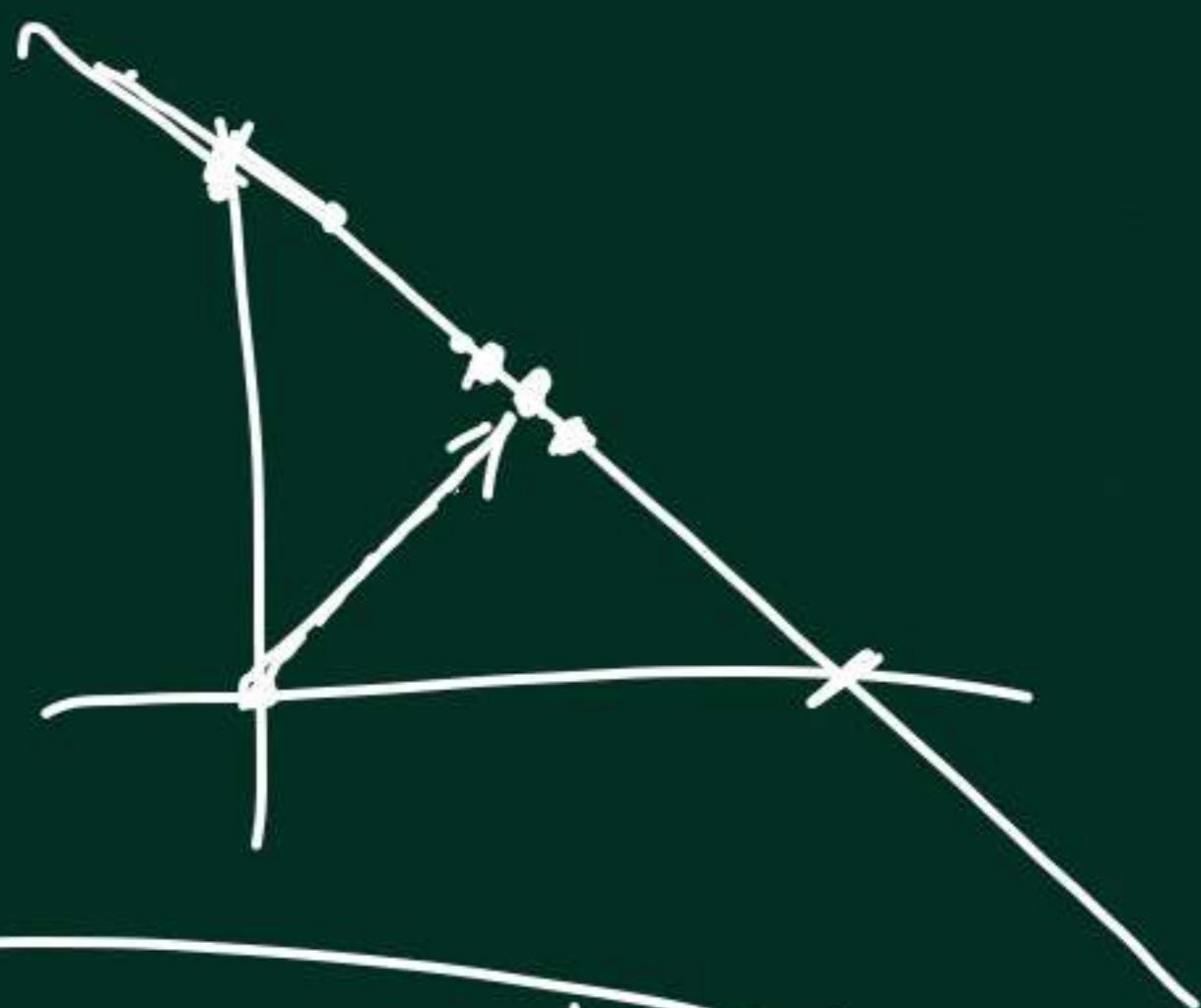
No. of words in line 1

— 1 — line 2

⋮  
⋮  
⋮

$i_1$   
 $i_1+1, i_2$   
 $i_2+1, i_3$   
 $\vdots$   
 $i_q$

exponentially  
many possible  
solutions.



## Greedy Ideas.

→ As many as words as possible

→ Average slack

Greedily try to be

close to average slack.)

} does  
this  
minimize  
sum

→ Check two lines at a time,  $(\text{slack})^2$ ?  
be greedy.

# Dynamic Programming

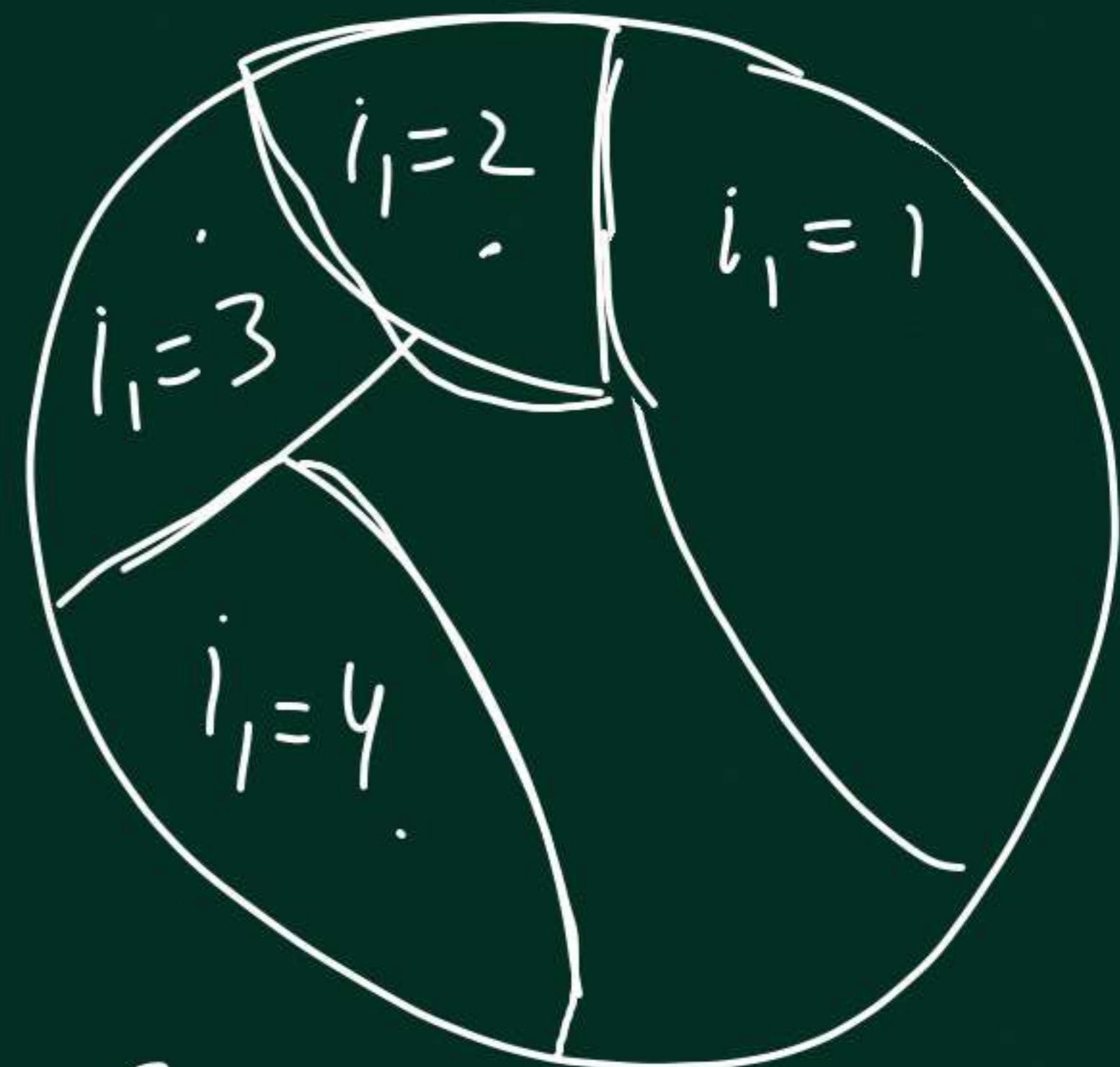
## Categorizing Possible Solutions

$i_1 \leftarrow$  last word in first line

$i_2 \leftarrow$  last - second

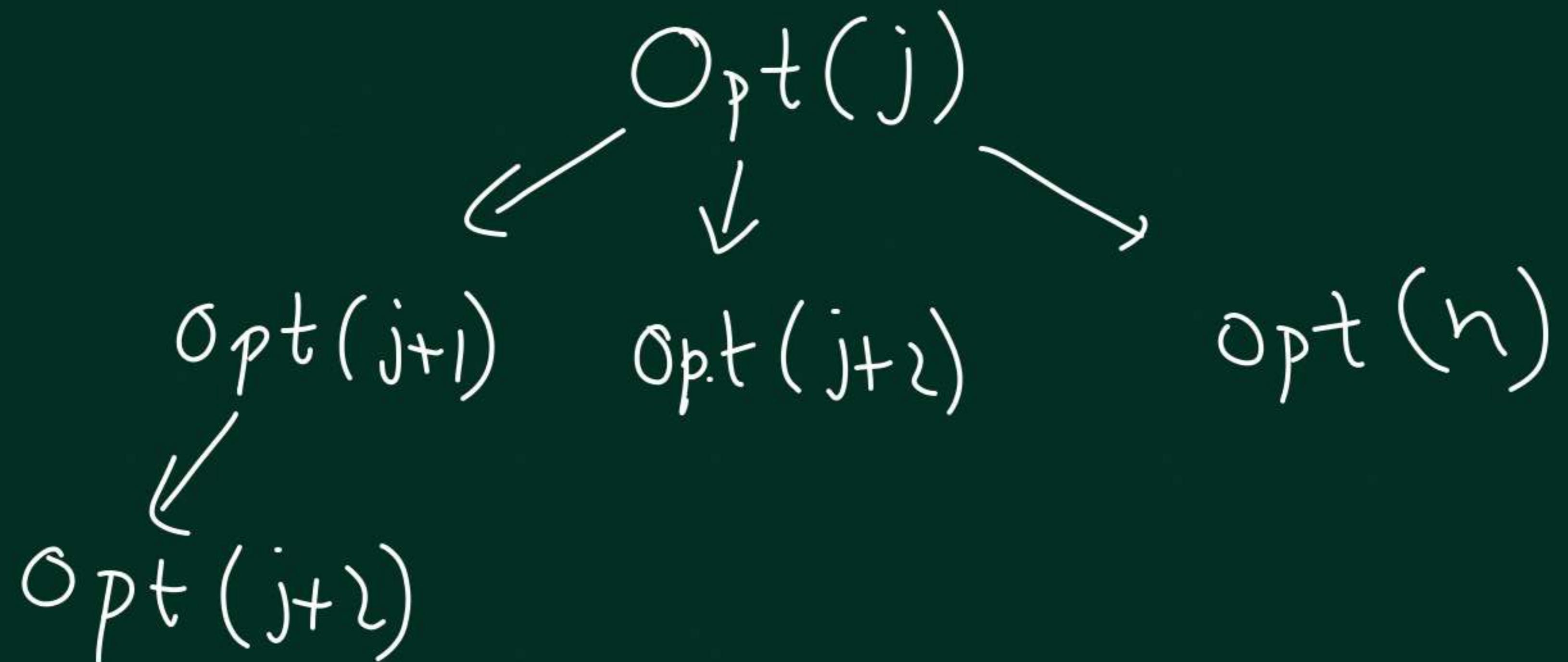
$i_3 \leftarrow$  last --- third.

$$OPT[1, n] = \min_{l_1} \begin{cases} (L - l_1)^2 + OPT[2, n] \\ (L - l_1 - l_2)^2 + OPT[3, n] \\ \vdots \\ (L - l_1 - l_2 - \dots - l_h)^2 + OPT[h+1, n] \end{cases}$$



No of "distinct" recursive calls.

$\text{Opt}(j) \leftarrow$  optimal value for the words  
 $l_j, \dots, l_n$



for ( j=n to 1 )

$$\text{opt}(j) = \min \left\{ \begin{array}{l} (L - l_j)^2 + \text{opt}(j+1) \\ (L - \underbrace{l_j - l_{j+1}}_{\vdots} - \dots - l_k)^2 + \text{opt}(j+2) \end{array} \right\}$$

Running time

$\Theta(n^2)$

Implementation

Optimal solution

Compute all slack squares

Before hand  $(L - l_j - l_{j+1} - \dots - l_k)^2$

# Edit Distance / Sequence Alignment

Distance between strings.

Spell checker — Most similar, closest

Training — Training

One way

4

SNOW

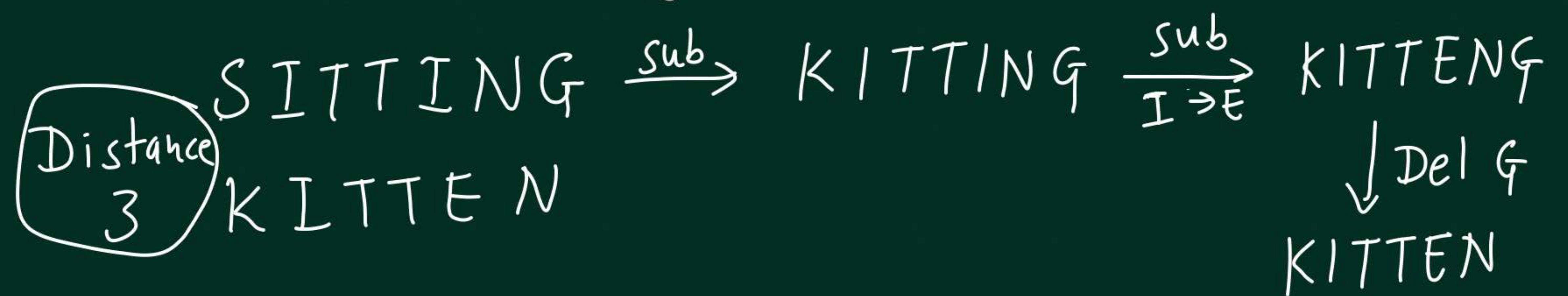
NOWN

TREE  
RACE

Hamming Distance. = no. of positions where you have diff characters

# Levenshtein Distance

= minimum no. of insertions, deletions  
or substitutions needed  
to go from one string to other



Deletion / Insertion ← cost 2

Substitution

(1) MEAN  
(2) NEAN  
2 NAN  
1 NAM  
NAME MEAN

VOWEL - VOWEL ← cost 1

Cons - Cons ← cost 1

VOWEL - Cons ← cost 3

NAME → MEA → ME → AME → NAME  
Cost 8

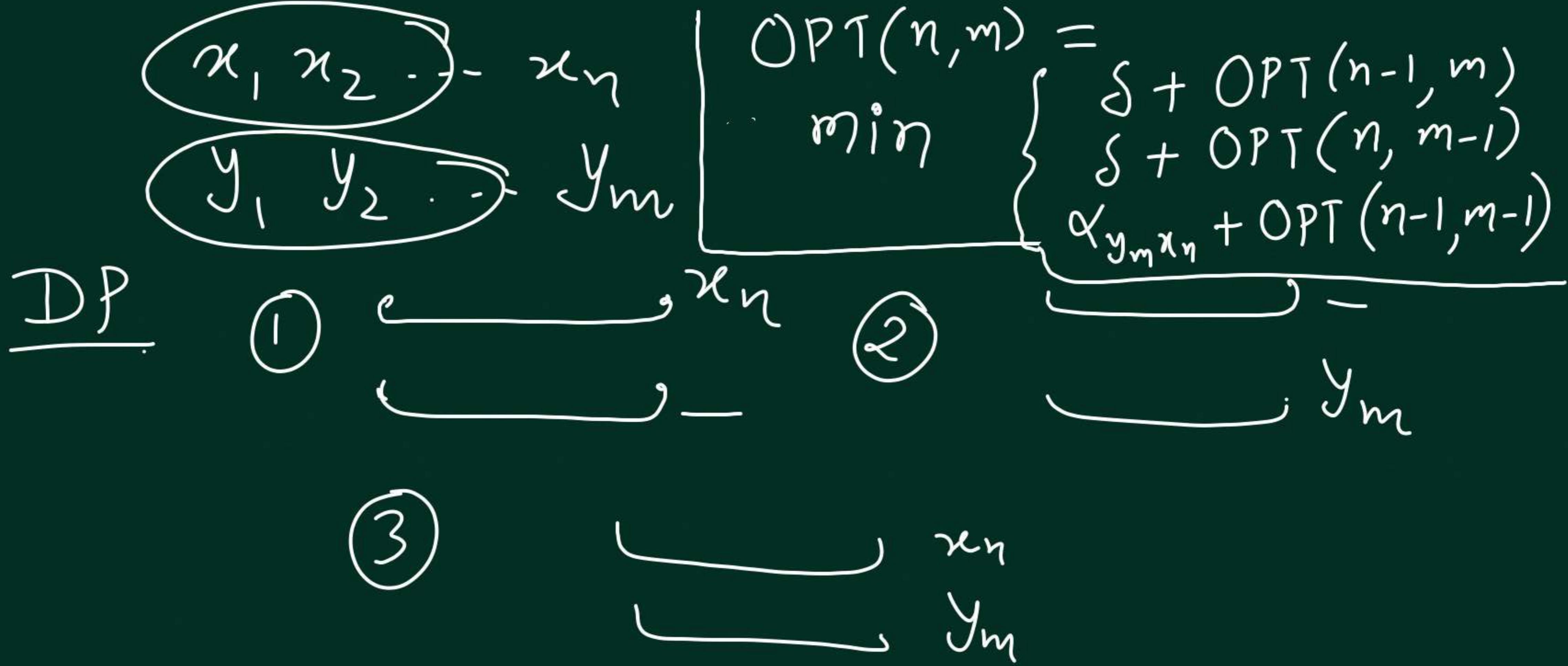
NAME

NEAN → NAAN → NAMN → NAME  
Cost 8

Sequence Alignment.

$\begin{matrix} \text{U} & \text{G} & \text{C} & \text{T} & \text{G} & \text{A} & \text{C} & \text{U} \\ \text{G} & \text{A} & \text{A} & \text{T} & \text{G} & \text{C} & \text{A} \end{matrix}$	$\delta$ $\alpha_{CA}$	$\delta$ $\delta_f$
$\begin{matrix} \text{U} & \text{(G)} & \text{C} & \text{T} & \text{(G)} & \text{A} & \text{C} & \text{U} \\ -\text{(G)} & \text{A} & \text{(T)} & \text{(G)} & \text{C} & \text{(A)} & - \\ - & - & - & - & - & - & - \end{matrix}$		

Penalties < Gap Penalty  $\delta$        $\alpha_{CA} + 4\delta$   
 < MisMatch Penalty



$\text{OPT}(n, m) \leftarrow$  optimal cost of Alignment  
 between  $x_1 \dots x_n$  and  $y_1 \dots y_m$

$OPT(i, j) \leftarrow$  opt cost for alignment  
between  $x_1 \dots x_i, y_1 \dots y_j$

$$OPT(i, 0) = i \cdot \delta$$

$$OPT(0, j) = j \cdot \delta$$

$$OPT(n, m)$$

for ( $i = 1$  to  $n$ )

for ( $j = 1$  to  $m$ )

$$OPT(i, j) = \min \left\{ \begin{array}{l} \delta + OPT(i-1, j) \\ \delta + OPT(i, j-1) \\ \alpha_{x_i y_j} + OPT(i-1, j-1) \end{array} \right.$$

	0	1	...	$n$
0	0	18	28	38
1	18			
2	28			
3	38			
$m$				

Optimal solution?

Opt cost < space  
 $O(\min(m, n))$

Space complexity =  $O(mn)$

Space  $O(m+n)$  ?  
 Time  $O(mn)$ .

Think about it

# Randomized Algorithms: 2-dim Linear Programming

Deterministic algorithms: same behaviour on a fixed input

Randomized algorithms: random decisions

→ running time → random variable

→ output can be random

- With high probability, running time is small
- w.h.p., Output is correct.

Randomized Quicksort (randomly choose pivot)

Sampling, Avoid worst cases

- With some prob., hardware can fail.
- Algorithms : repeat multiple times  
⇒ error probability very small

Many cases: randomized algorithms are the fastest known.

Also, simplest.

2-dim

# Linear Programming

$x, y$

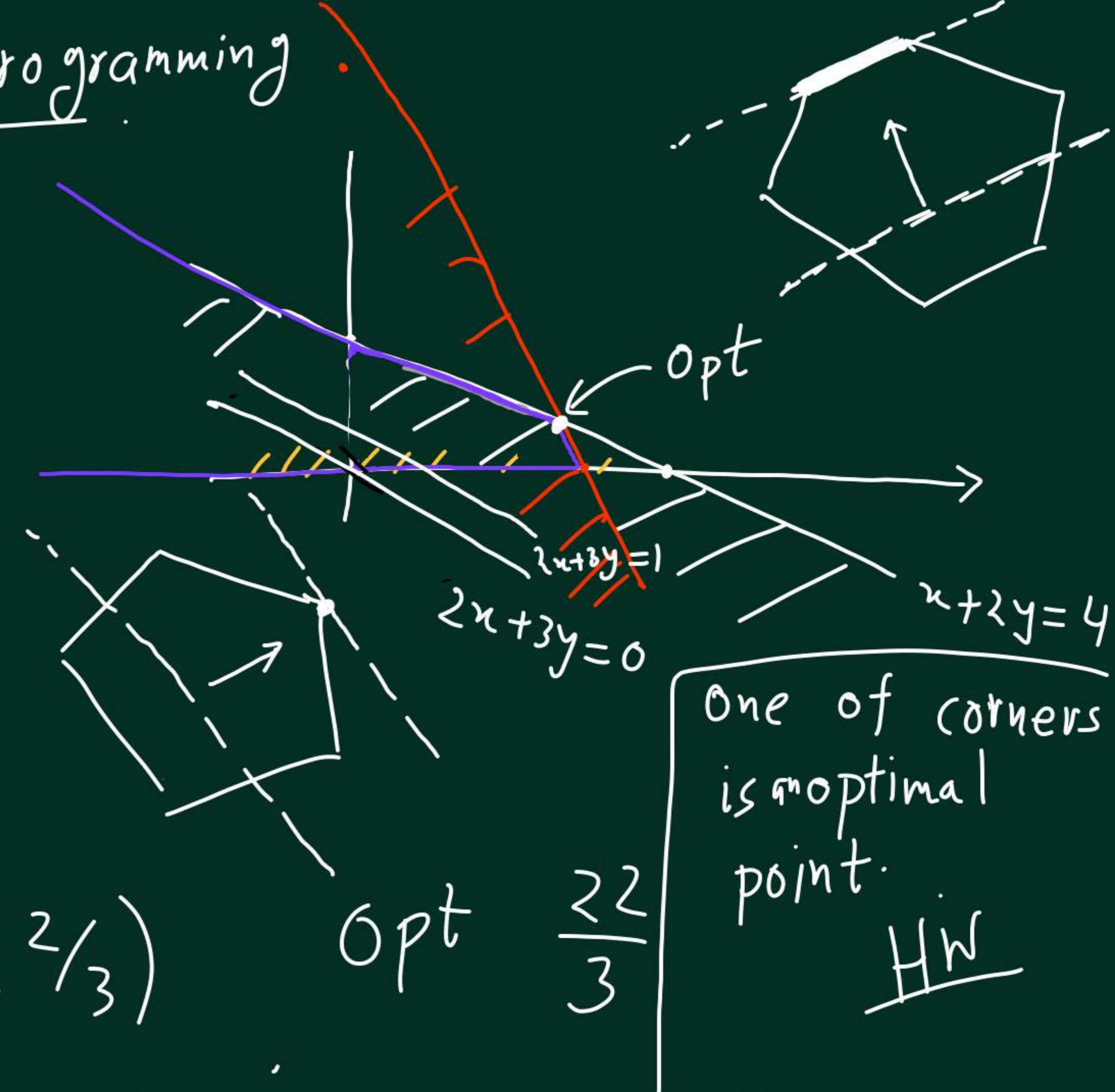
$$x + 2y \leq 4$$

$$2x + y \leq 6$$

$$y \geq 0$$

$$\max 2x + 3y$$

$$(8/3, 2/3)$$

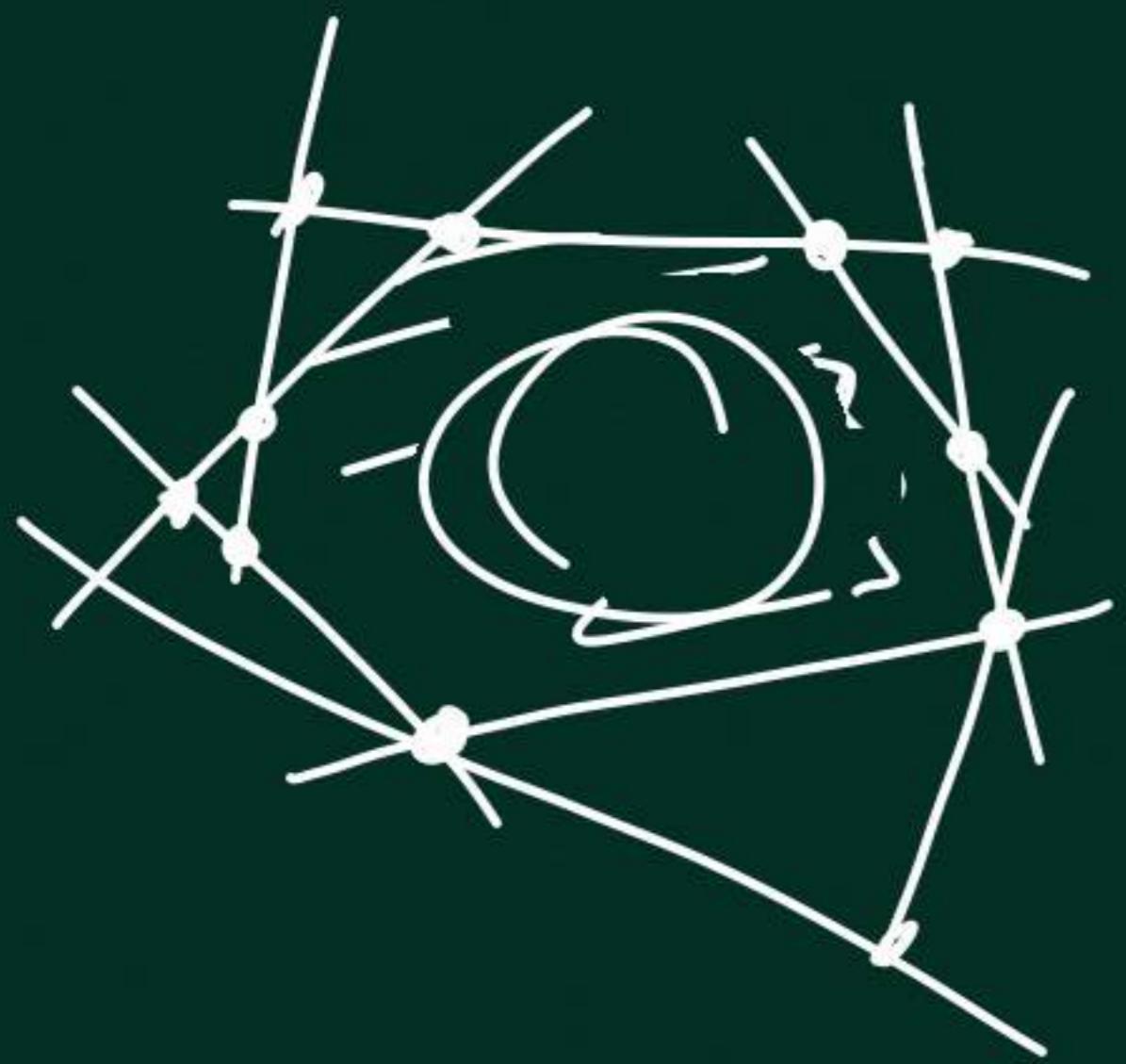


$n$  linear constraints.

check all the corners

$O(n^3)$

at most  $n$



no. of intersection  
points  
 $O(n^2)$



~~Hw~~  $O(n \log n)$  compute all corners.

Randomization

Find the optimal point

simple algorithm

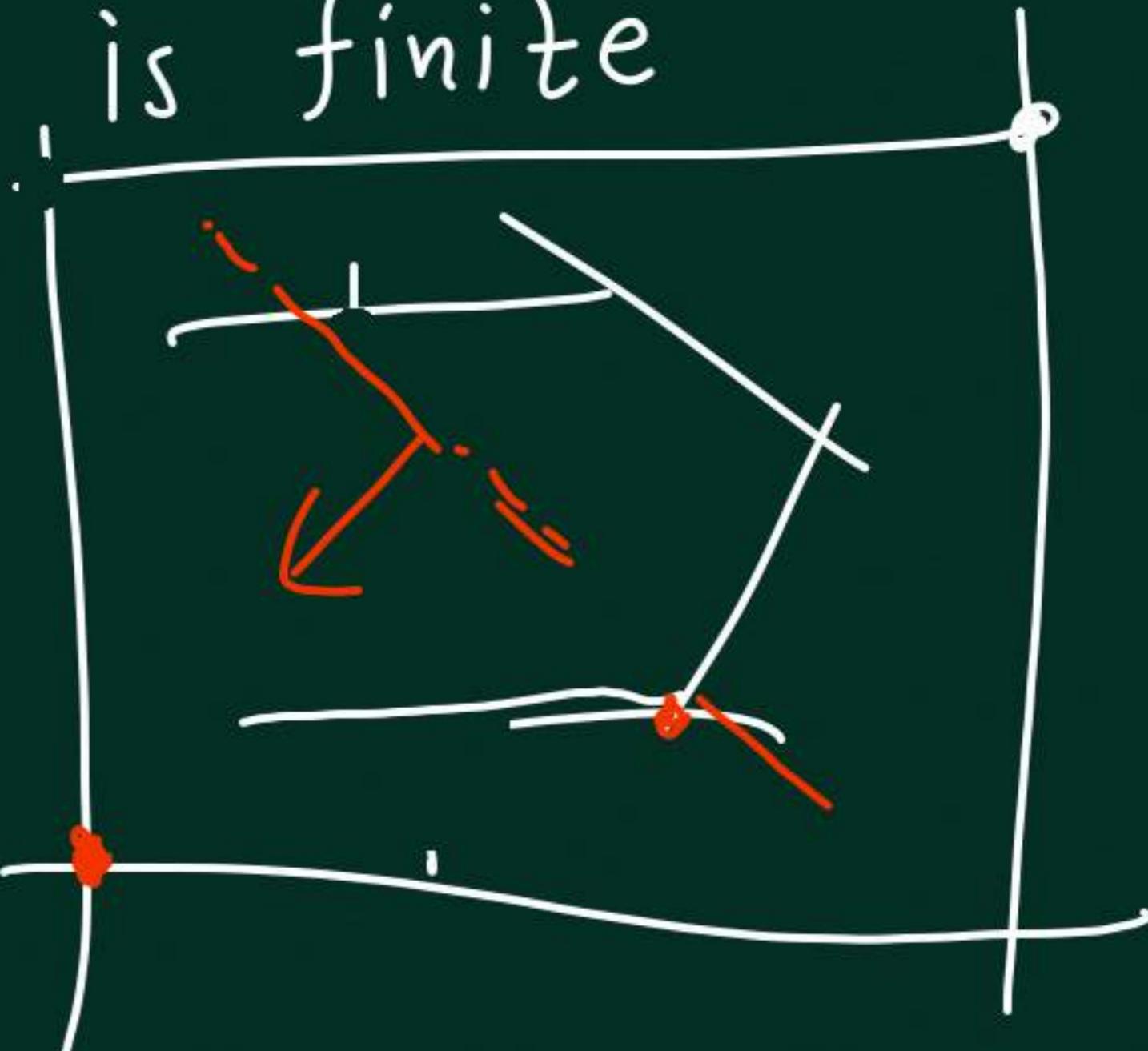
$O(n)$  time ( $w \cdot g \cdot p$ )

Assumptions: optimal value is finite

for every intersection point

$$-L < x\text{-coord} < L$$

$$-L < y\text{-coord} < L$$



Assumption : optimal solution is unique.

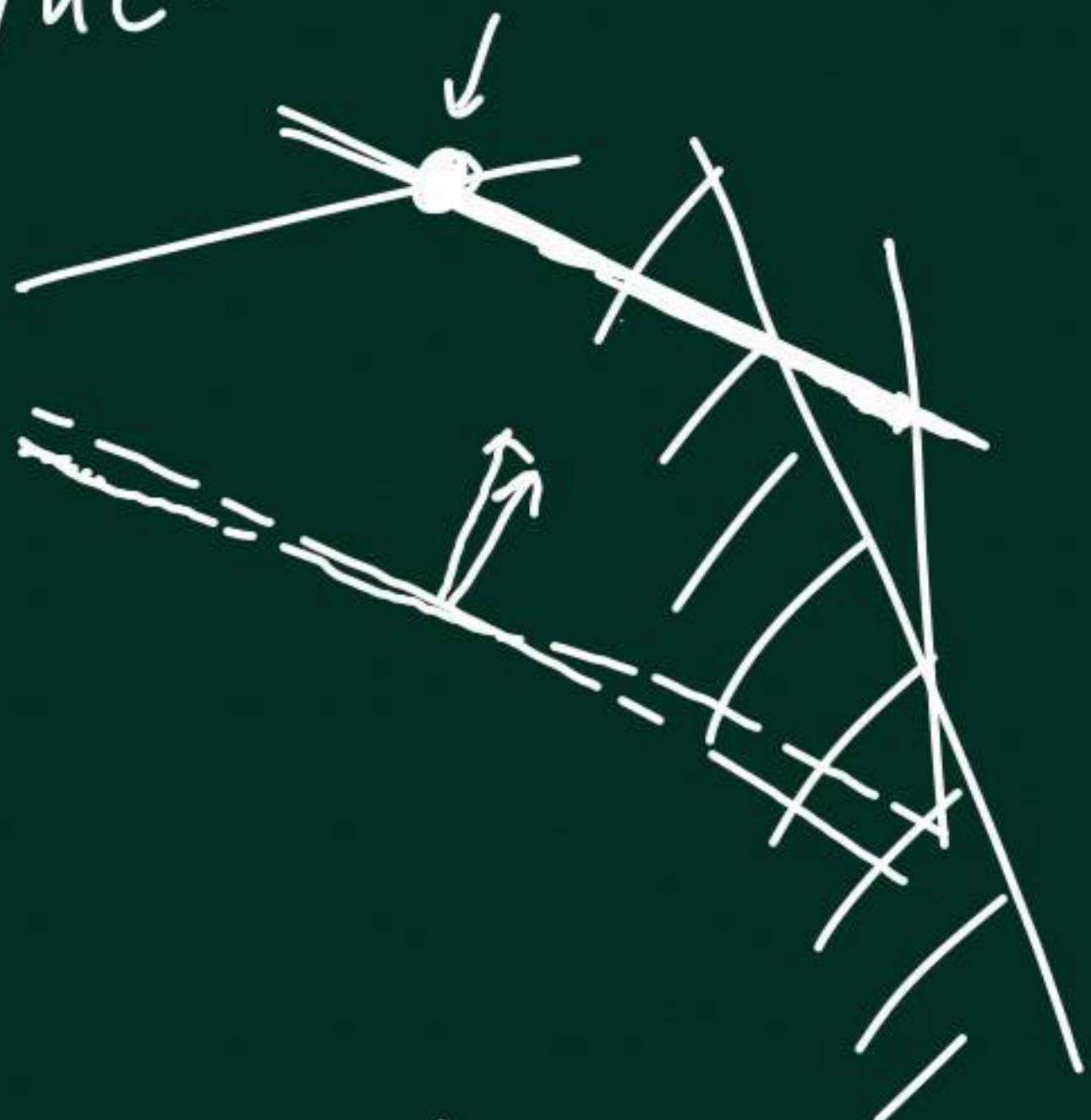
$C_0 \leftarrow$  Set of constraints  $\leftarrow -L \leq x \leq L$   
 $-L \leq y \leq L$

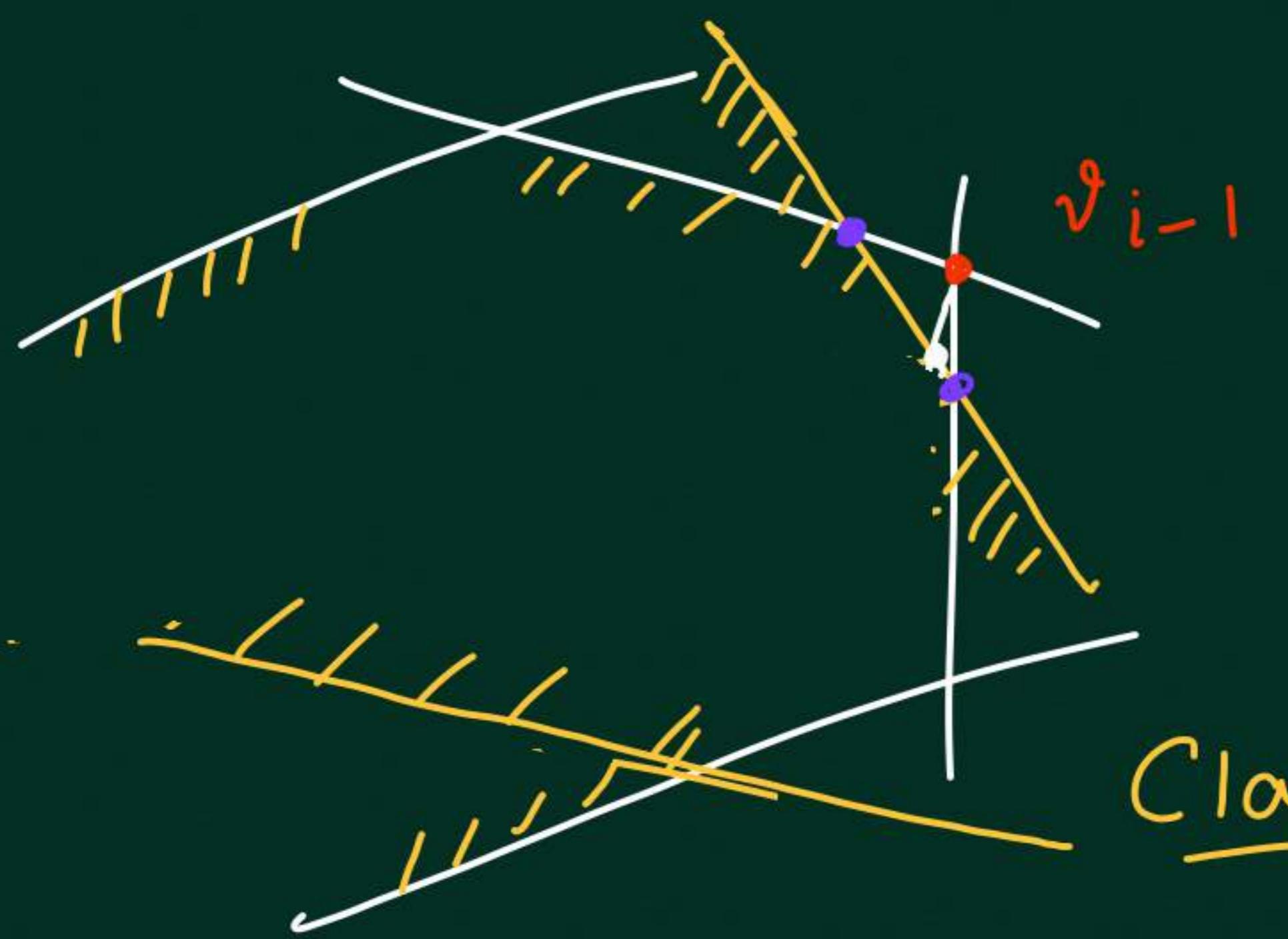
$v_0 \leftarrow$  Optimal point  $\leftarrow (L, L)$

Add constraints one by one and  
update the optimal point

$(C_{i-1}, v_{i-1}) \longrightarrow (C_i, v_i)$

$$C_i = C_{i-1} \cup \{h_i\}$$





If  $v_{i-1}$  satisfies  $h_i$   
then  $v_i = v_{i-1}$   $O(1)$

If  $v_{i-1}$  doesn't satisfy  $h_i$   
 $v_i$  will satisfy  $h_i$   
with equality  $O(i)$

## Maximization

$$f(p) = c_p x + d_p y$$

objective function

## Obs

$$f(v_{i-1}) \geq f(v_i)$$

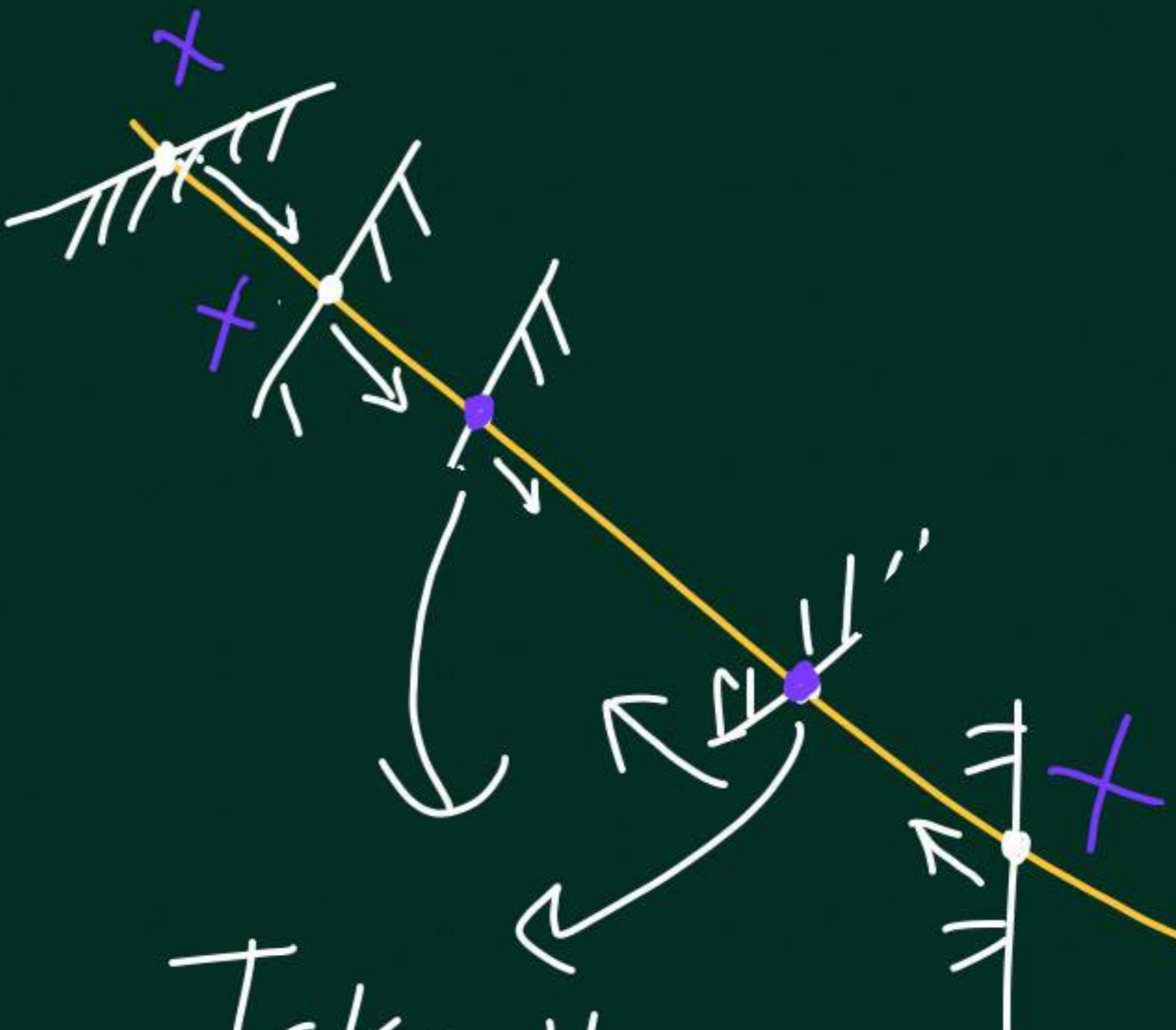
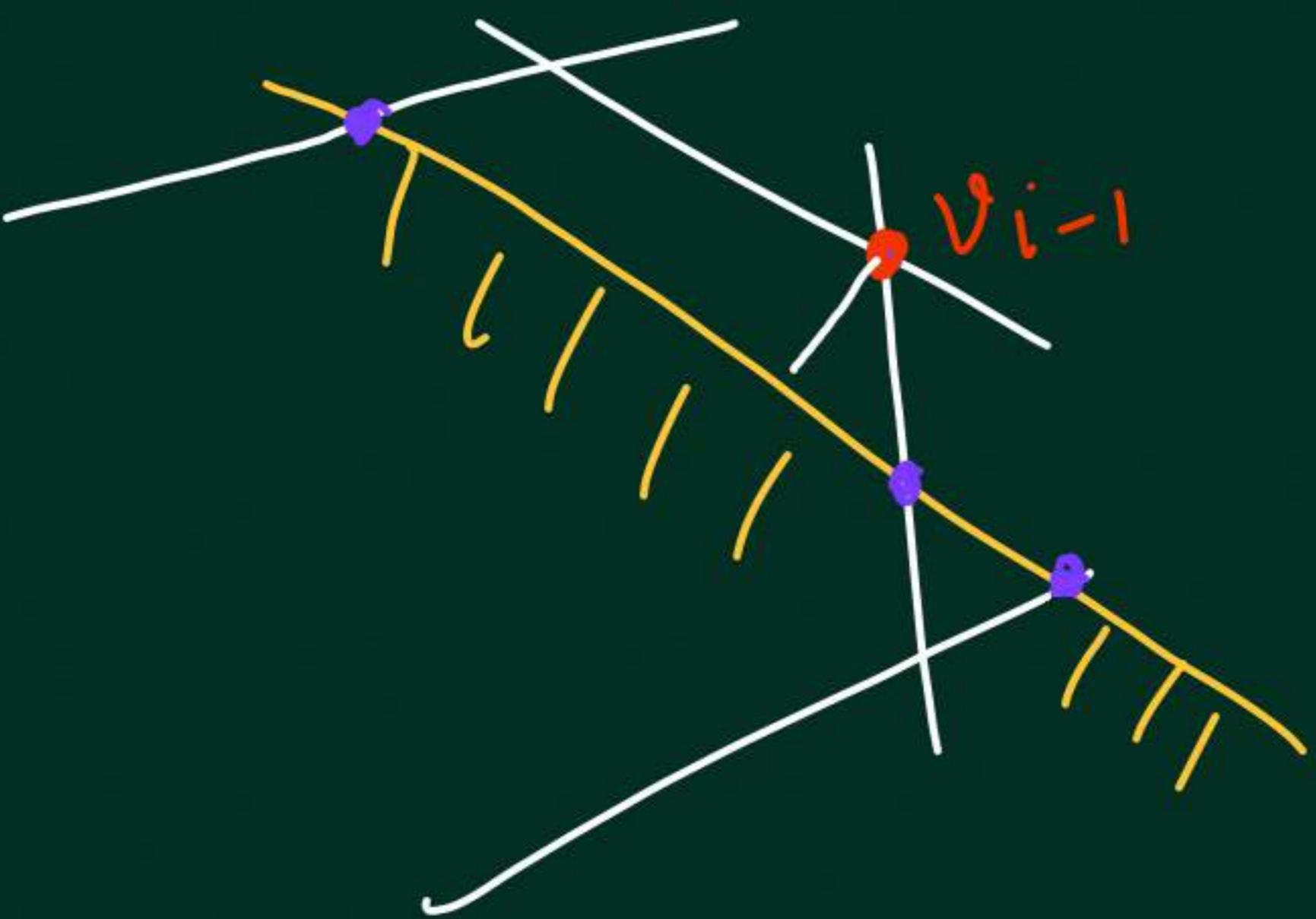
Consider line segment joining  
 $v_i$  and  $v_{i-1}$ .  
Intersection of line segment  
with  $h_i$  is q.

$$f(v_{i-1}) \geq f(q) \geq f(v_i)$$

$O(i)$  time  
update algorithm?

Obs  $v_i$  is the point of intersection of  $h_i$  and

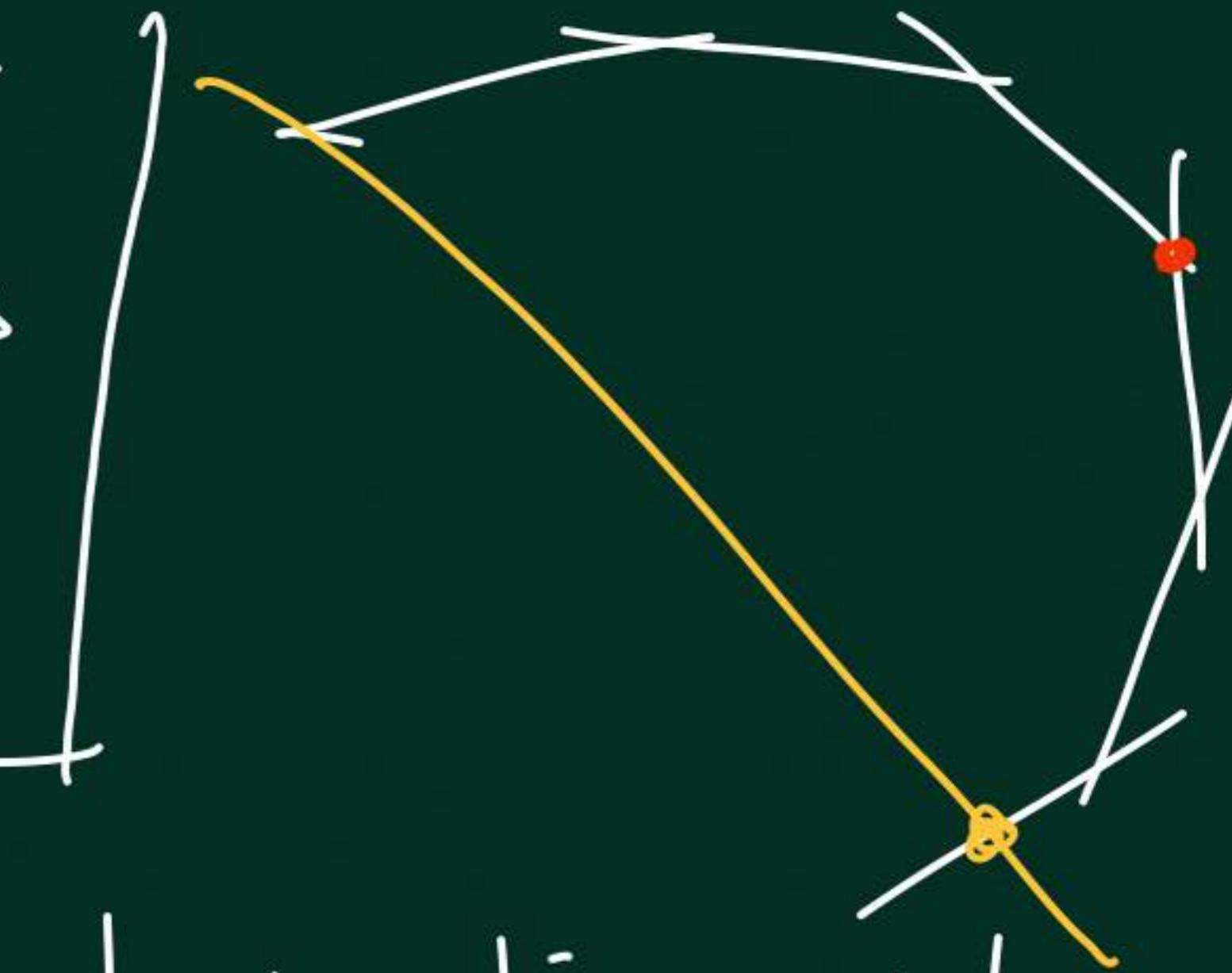
Some  $h_l$   $1 \leq l \leq i-1$ .



Take the best of two  $\leftarrow v_i$



Also detect if  
System becomes  
infeasible.

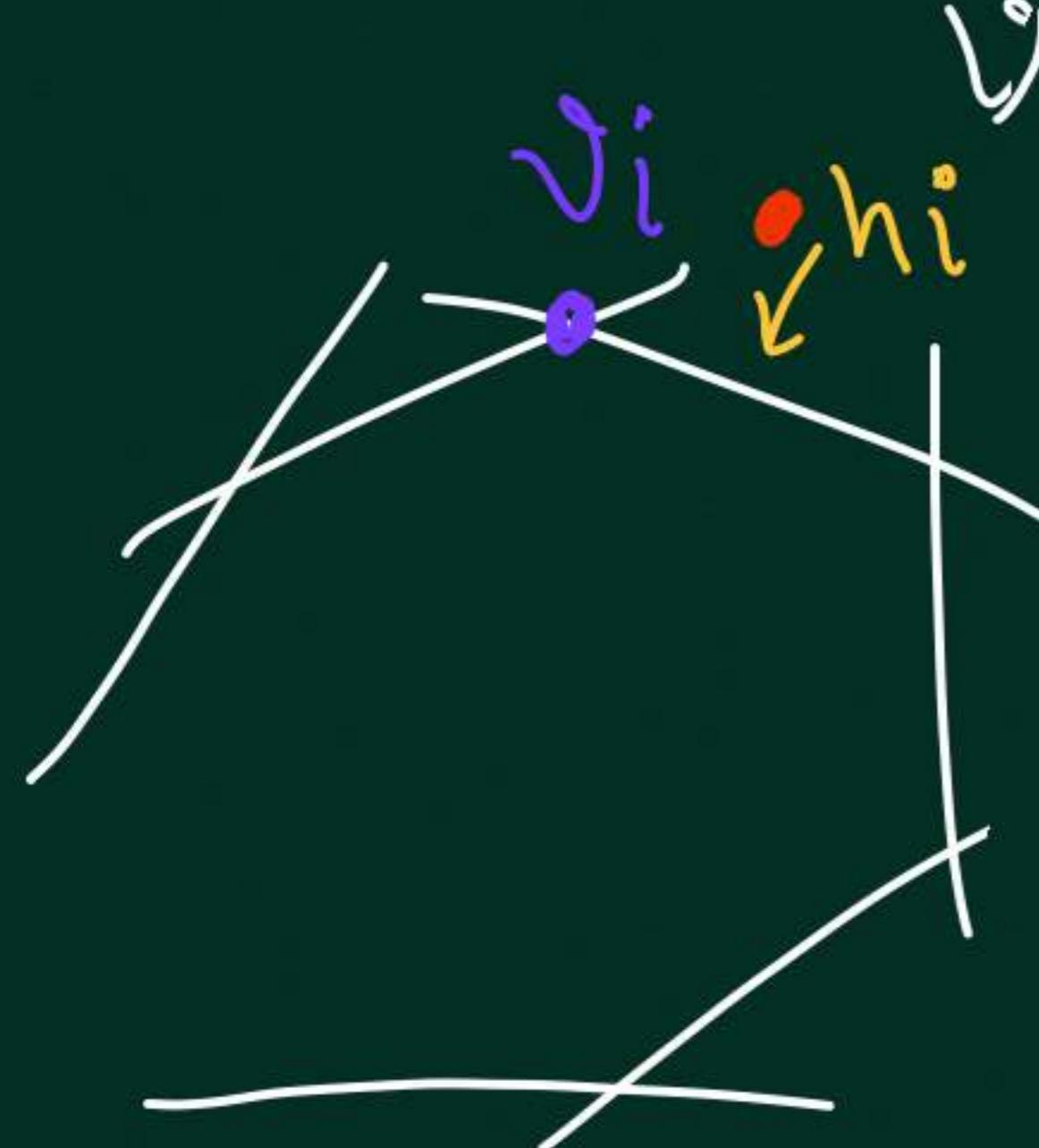


$i^{\text{th}}$  iteration  $\leftarrow$   $i-1$  intersection point

$$\begin{aligned}\text{Total running time} &= O(1) + O(2) + \dots + O(i) + \dots + O(n) \\ &= O(n^2)\end{aligned}$$

We arrange the lines in a random order  
and then run the algorithm.

Que What is the probability that  
 $v_i \neq v_{i-1}$  (bad event)



$$= \text{prob that } i^{\text{th}} \text{ line is} \\ \text{one of two lines passing through } v_i \\ = \frac{2}{i}$$

Expected number of times i have  
computed intersections of lines.

$X_i$  = the no. of intersections  
computed in the  $i^{\text{th}}$  iteration.

Expected  
running

time

$O(n)$

$$X_i = \begin{cases} 0 & \text{with pr } 1 - \frac{2}{i} \\ i-1 & \text{pr } 2/i \end{cases}$$

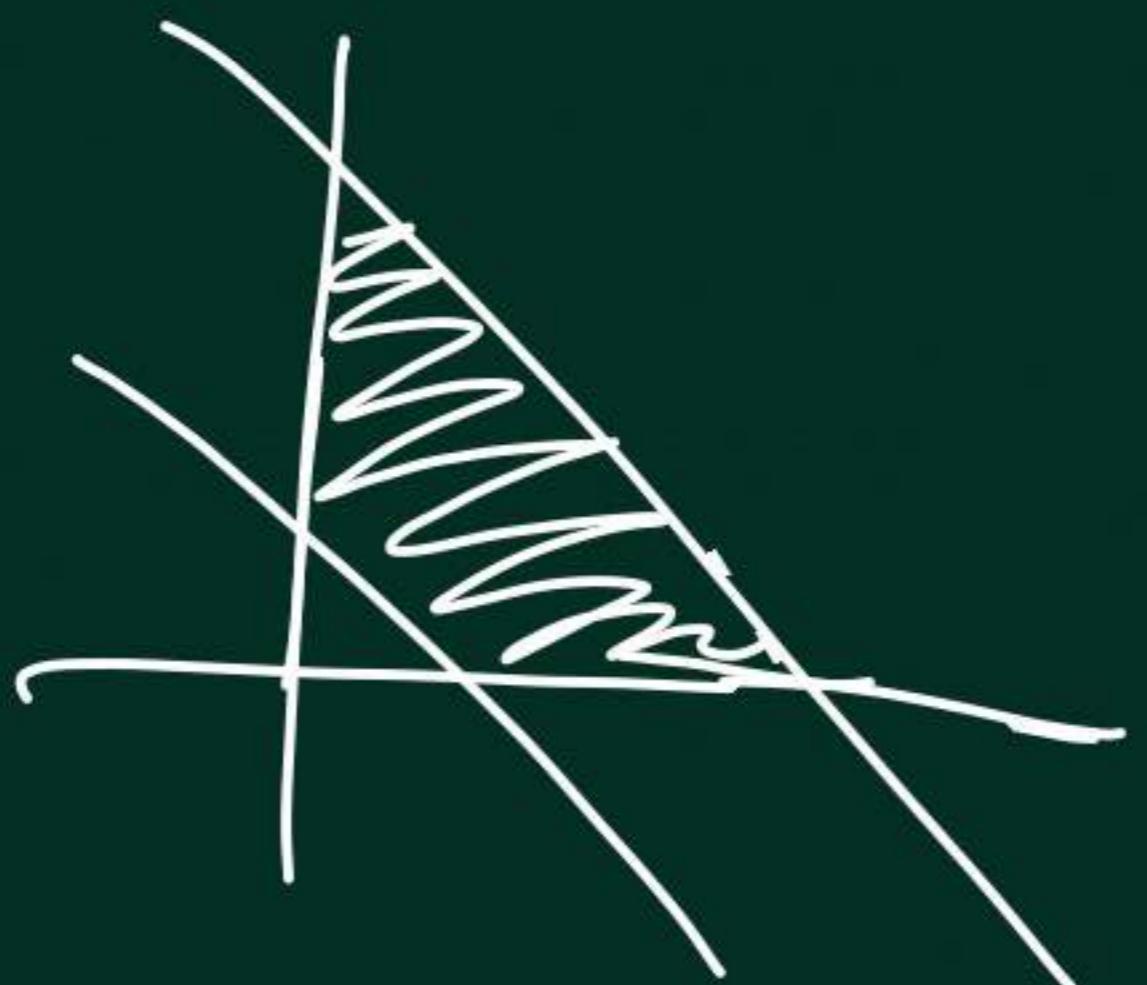
$$\leq 2n$$

$$X = \sum_{i=1}^n X_i \quad E[X] = \sum_{i=1}^n E[X_i] = \sum_i \frac{2(i-1)}{i}$$

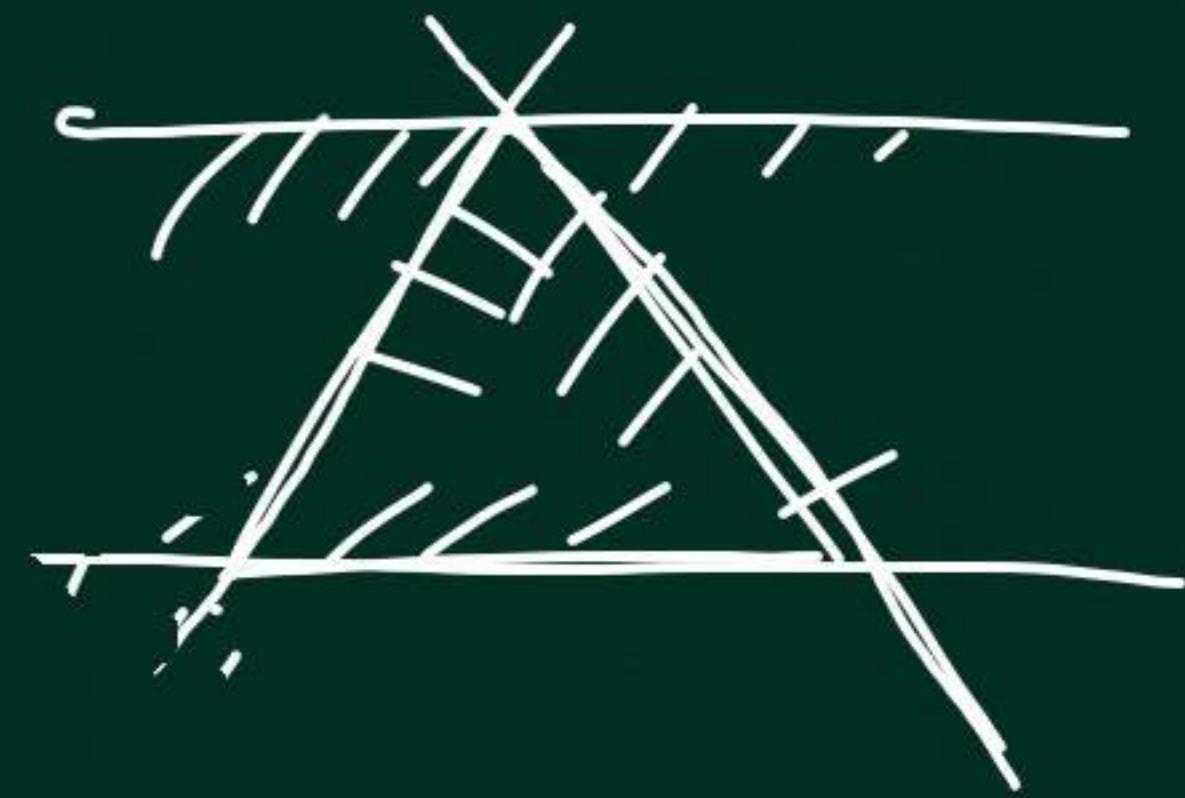
# Simplex Algorithm (not known to be in P)

Recall: one of the corners is an optimal so

$$\begin{cases} 2x + 3y \leq 5 \\ x + y \geq 1 \\ x \geq 0 \\ y \geq 0 \end{cases}$$



$$x + y + z \geq 1$$



Convex set  
Polyhedron / Polytope



Def (corner): For a system of linear inequalities in  $n$  variables, a corner is a feasible point which satisfies  $n$  linearly independent constraints with equality.

Def A corner is a feasible point s.t. there exist a hyperplane  $H$  s.t.  $H \cap$  polyhedron is a unique point

## Simplex

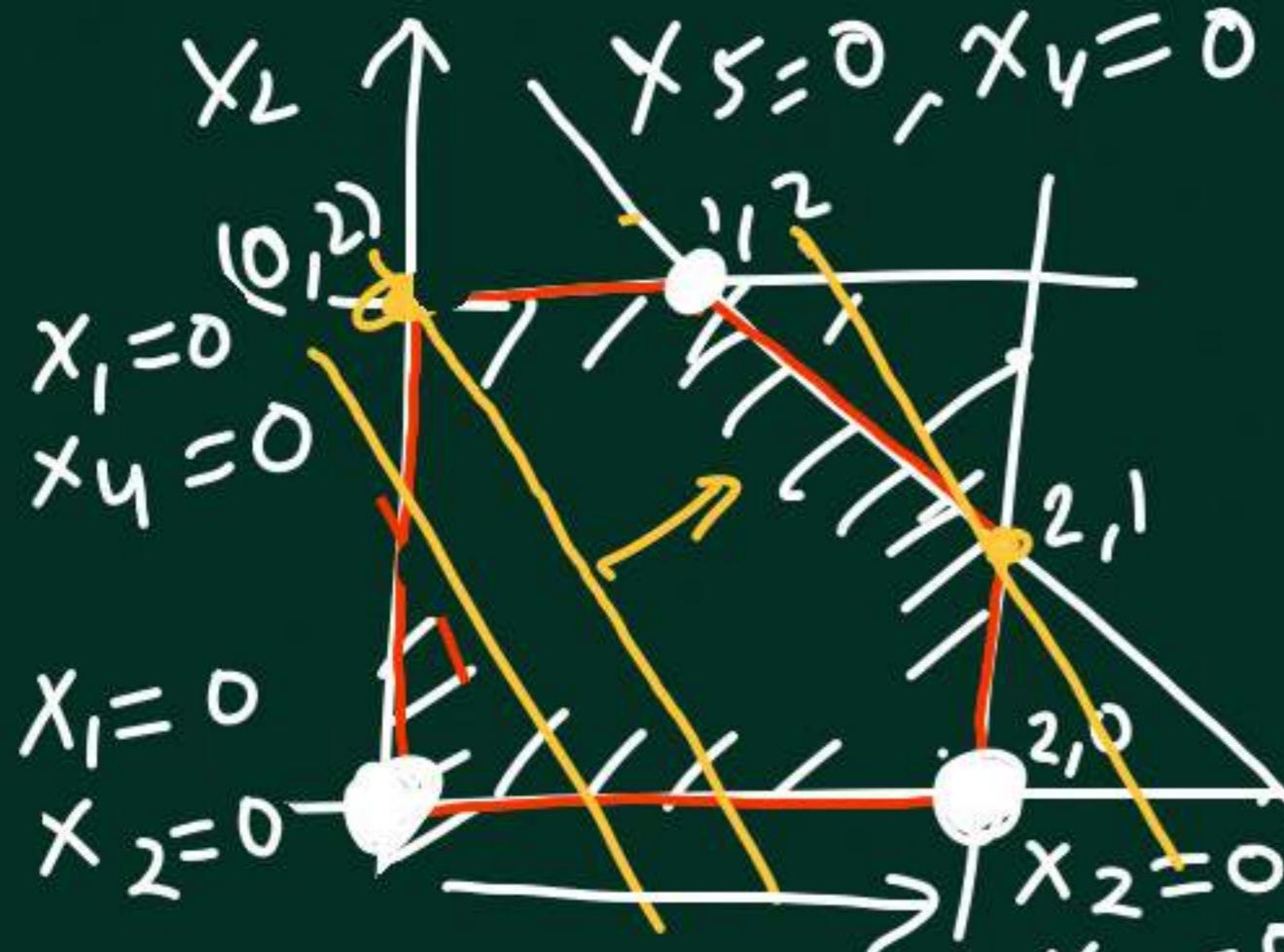
- Start from a corner
- keep going to a neighboring corner such that the objective function increases.
- Stop when there is no neighboring corner with higher objective value.

$n$ -dim hypercube

$$\begin{aligned} 0 \leq x_1 &\leq 1 \\ 0 \leq x_2 &\leq 1 \\ &\vdots \\ 0 \leq x_n &\leq 1 \end{aligned} \left\{ \begin{array}{l} 2^n \text{ corners} \end{array} \right.$$

Hirsch Conjecture:

In  $n$ -dimensional polytope  
with  $m$  constraints  
there is always a  
path of length  $\text{poly}(n, m)$   
between any two corners.



Initial basic feasible solution  
Corner

Assume: we have an initial BFS  
 $\left\{ \begin{array}{l} x_1 \geq 0 \\ x_2 \geq 0 \end{array} \right\}$  Non-negativity constraints. Standard Forms of linear Program

$$x_1 \geq 0, x_2 \geq 0$$

$$x_1 \leq 2$$

$$x_1 = 2 - x_3, x_3 \geq 0$$

$$x_2 \leq 2$$

$$x_2 = 2 - x_4, x_4 \geq 0$$

$$x_1 + x_2 \leq 3$$

$$x_1 + x_2 = 3 - x_5, x_5 \geq 0$$

$$\text{Max } 2x_1 + x_2$$

Start with

$$x_1=0, x_2=0 \quad x_3=2 \quad x_4=2 \quad x_5=3$$

∴ Obj =  $2x_1 + x_2$

Allow one variable  
to increase

Let me choose to increase  $x_1$

$x_2$  will remain zero.

$$x_3 \geq 0 \Rightarrow 2 - x_1 \geq 0 \Rightarrow x_1 \leq 2$$

$$x_5 \geq 0 \Rightarrow 3 - x_1 - x_2 \geq 0 \Rightarrow x_1 \leq 3$$

$$\begin{array}{ll} x_1 = 2 & x_3 = 0 \\ \underline{x_2 = 0} & x_4 = 2 \\ & x_5 = 1 \end{array}$$

Rewrite the obj  
function in terms  
of  $x_2, x_3$

$$\begin{aligned} \text{Obj} &= 2(2-x_3) + x_2 \\ &= 4 - 2x_3 + x_2 \end{aligned}$$

Should increase  $x_2$

$x_3$  will remain  
zero

$$2 - x_2 \geq 0$$

$$x_2 \leq 2$$

$$x_1 + x_2 \leq 3$$

$$2 - x_3 + x_2 \leq 3$$

~~$$x_2 \leq 1 + x_3$$~~

change  $x_2 \leftarrow 1$

$$x_1 = 2 \quad x_4 = 1$$

$$x_2 = 1 \quad x_5 = 0$$

$$x_3 = 0$$

rewrite obj  
in terms

$$x_3, x_5$$

$$\text{Obj } 4 - 2x_3 + \cancel{- x_5}$$

$$\begin{aligned} \text{Obj} &= 4 - 2x_3 + x_2 \\ &= 4 - 2x_3 + (3 - x_1 - x_5) \\ &= 7 - 2x_3 - x_1 - x_5 \\ &= 7 - 2x_3 - (2 - x_3) - x_5 \\ &= 5 - x_3 - x_5 \end{aligned}$$

The algorithm stops here because both  $x_3$  and  $x_5$  have negative coefficients in the objective function. Hence, we cannot increase any of the two. The optimal value is 5 (as  $x_3=x_5=0$ )