

**DP 5** Discrete Knapsack items (value, weights)

Whole or nothing

$v_1 v_2 \dots v_n$   
 $w_1 w_2 \dots w_n$

$Z =$  Knapsack total weight

Obj: max total value  
 $\sum \text{weights} \leq Z$

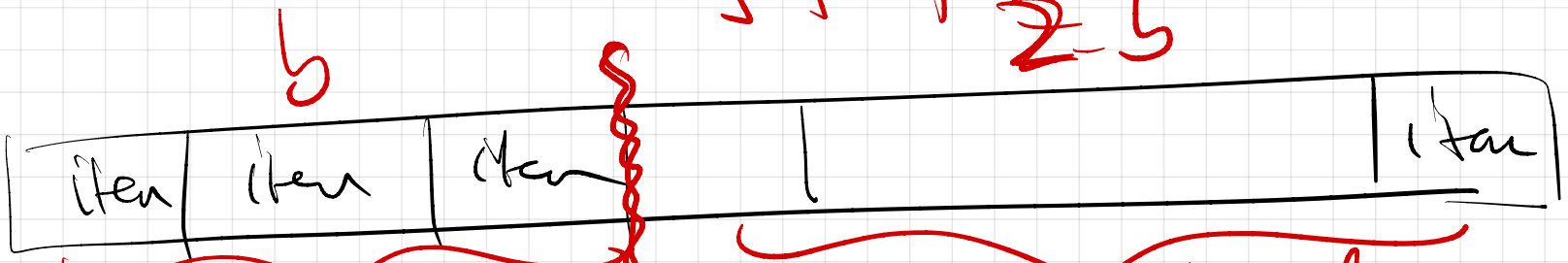
• Critical diff vs coins, rod-cuts:

table changes because each item can be used 0/1 times.

itemset (or similar) must be part of changing input  $Z-b$

①

Knapsack



optimal  $[b; \text{all items}]$

optimal  $[Z-b; \text{all items not on left}]$



Indexing Trick: index elements global order  
 1, 2, 3, ..., n not changeable

not necess. input order, weight-sort, value-sort

Item set  $I_n = \{1, 2, \dots, n\}$   $I[1:n]$   
 Partial sets  $I_{n-1} = \{1, 2, \dots, n-1\}$   $I[1:n-1]$   
 $I_{n-2} = \{1, 2, \dots, n-2\}$   $I[1:n-2]$   
 $I_4 = \{1, 2, 3, 4\}$   $I[1:4]$

}  $\Theta(n)$   
 subsets  
 (prefix)

step 1?  
 step 2A

Knapsack items available

Item chosen

prefix set

$$C[Z, I_n] = \max \left( \underset{\times_n}{V_n} + C[Z - w_{\times_n}, I_{n-1}] \right)$$

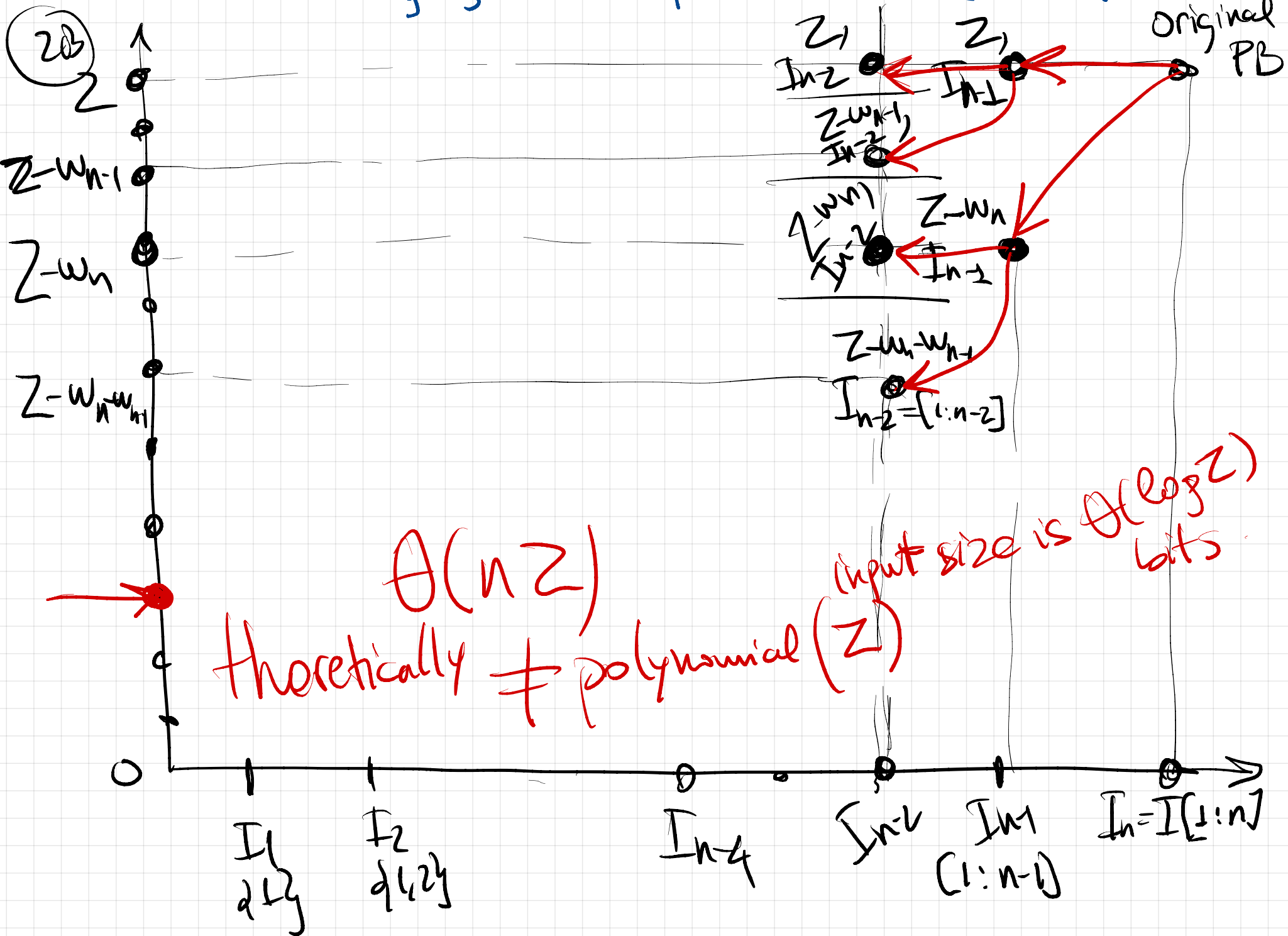
Search (max)

don't use item n

$$S[Z, I_n] = \begin{cases} 1 & \text{if } n \text{ is used} \\ 0 & \text{if not} \end{cases}$$

$\Theta(1)$

Wishful thinking: use prefix subsets (indexing trick)



$\Theta(nZ)$   
 theoretically  $\neq$  polynomial ( $Z$ )  
 (input size is  $\Theta(\log Z)$  bits)

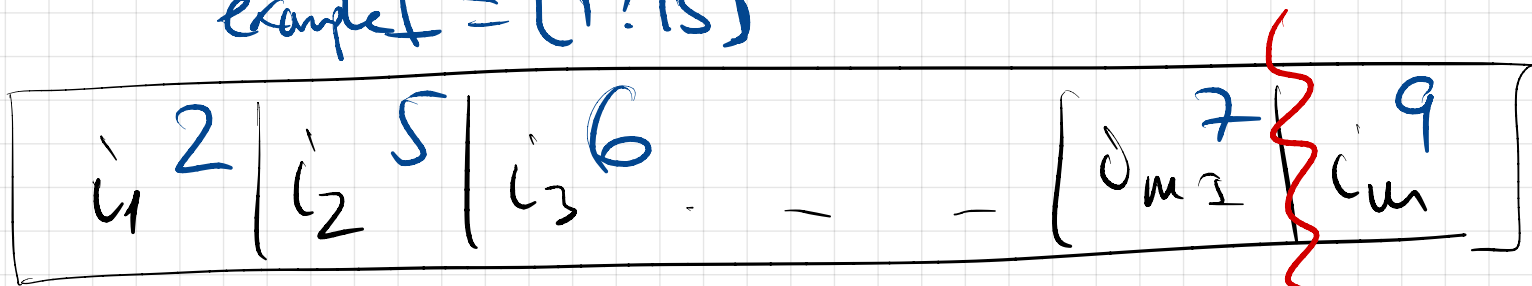
Y axis  $Z, Z-w_n, Z-w_n-w_{n-1}, \dots, Z-w_{n-1}$

$Z$  - subset of weights

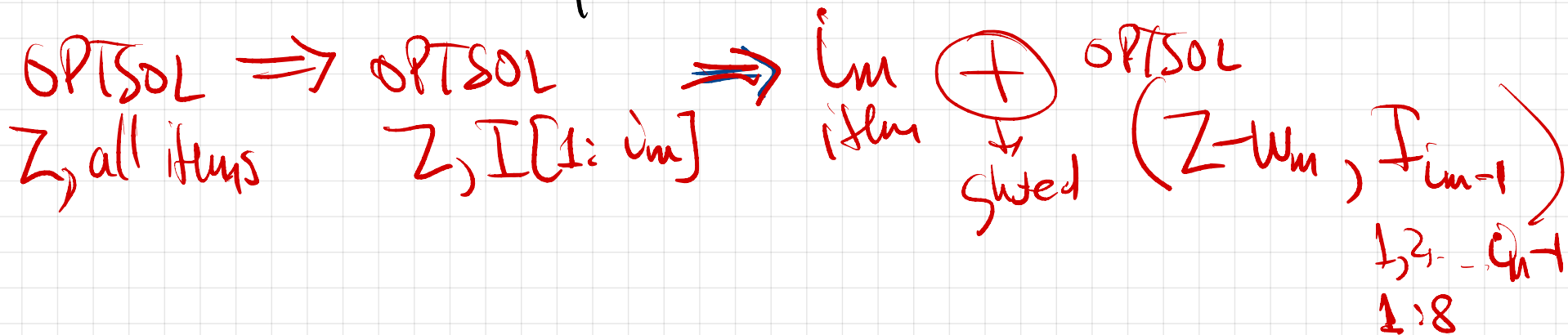
• requirement:  $Z, w$  integers (discrete range)

example  $I = [1:15]$

Step 1:  
OPTSOL



assume OPTSOL its presented in <sup>global</sup> index order



# Matrix Chain Multiplication

ex  $A \times (B \times C)$  =  $(A \times B) \times C$

$n=3$   $3 \times 5$   $5 \times 10$   $10 \times 20$  =  $3 \times 5$   $5 \times 10$   $10 \times 20$

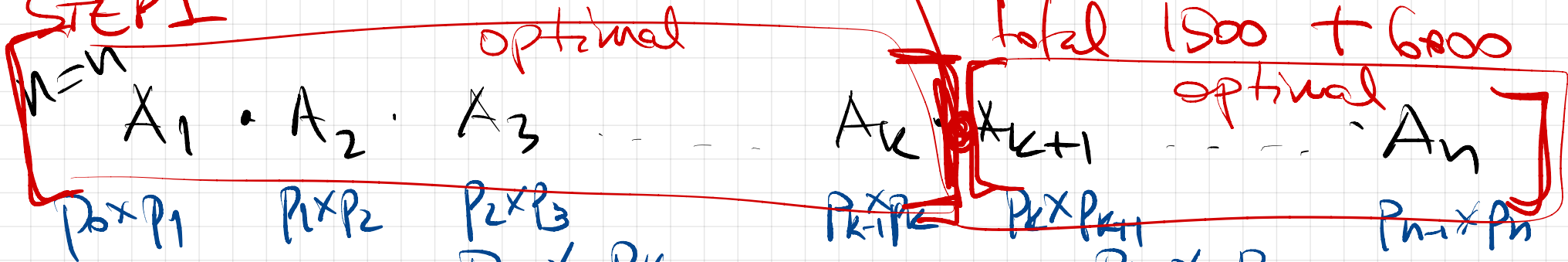
cell

$5 \times 10 \times 20$   
 $\Downarrow$   
 $5 \times 20$   
 $3 \times 5 \times 20$

$3 \times 5 \times 10$   
 $\Downarrow$   
 $3 \times 10$

total  $3000 + 1000$

last operation  
 total  $1500 + 6000$



What is the best multipl order?  $\equiv$  putting parentheses

last multipl (op) :  $p_0 \times p_k \times p_n$

Brute force: Try all possible parenthesis

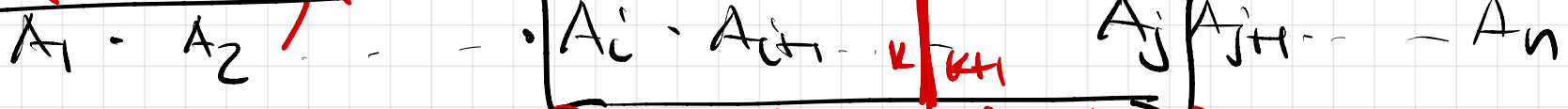
$n=4$

$(A(B(CD)))$  |  $((AB)C)D$  |  $(A(BC))D$  |  $A((BC)D)$  |  $A(B(CD))$

#ways to parenthesis = Catalan(n)  $\approx 4^n$  exact  $\binom{2n}{n} - \binom{2n}{n-1}$

$(A((BC)D))(EF)$

Step 2



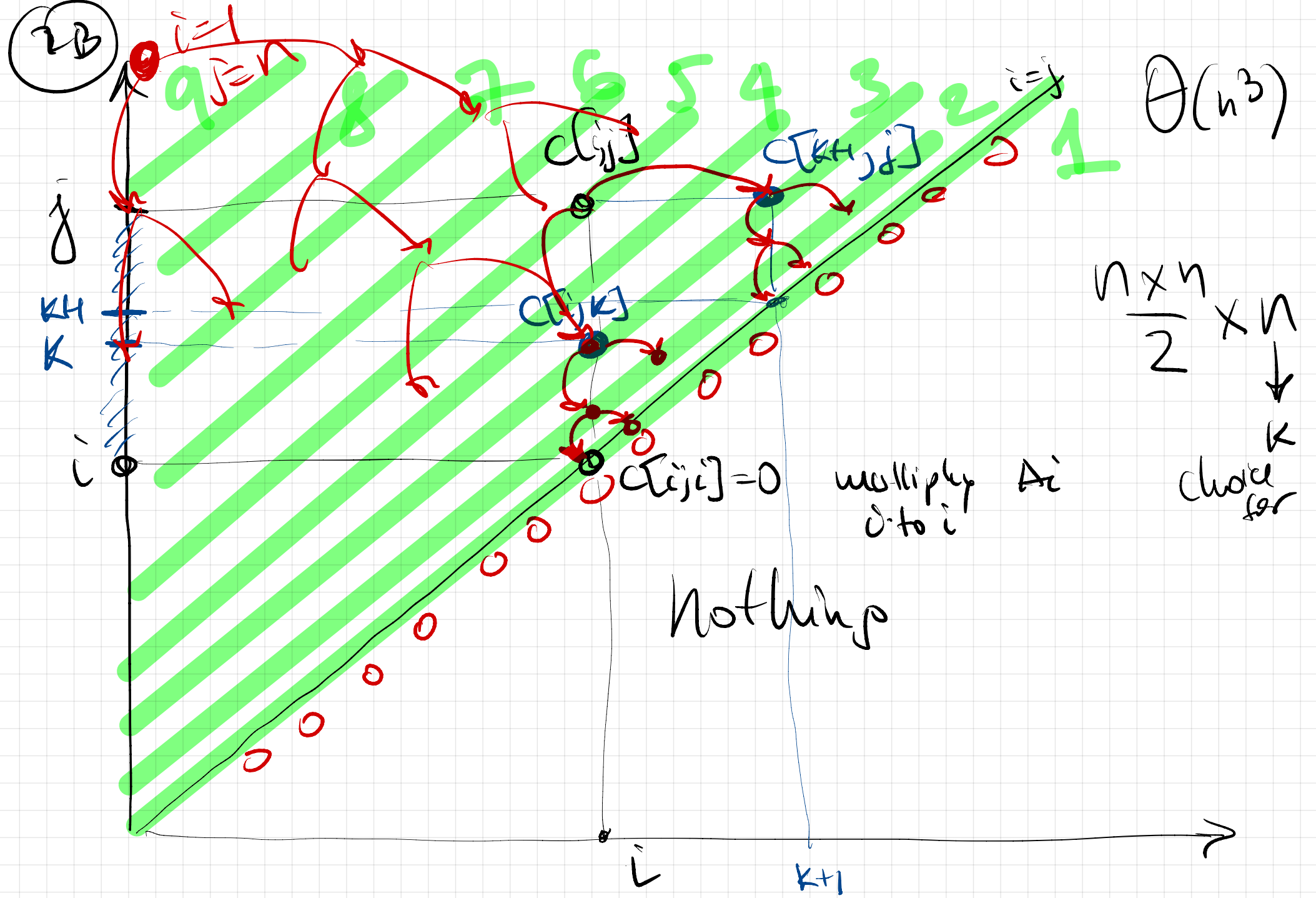
$C[i, j]$  = best cost for multiplying  $A_i \cdot A_{i+1} \cdot \dots \cdot A_j$

cost last op

$$p_{i-1} \cdot p_k \cdot p_j + C[i, k] + C[k+1, j]$$

$$i < k < j$$

$$S[i, j] = k$$



2B

$\Theta(n^3)$

$n \times n$   
 $\frac{1}{2} \times n$   
 $\downarrow$   
 choice for k

Nothing

multiply 0 to i

$C[k][j]$

$C[k][i]$

$C[i][j]$

$i$

$k+1$

$k$

$j$

$i=1$   
 $j=n$

$i=1$

$C[i][i]=0$

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0

0



MEMOIZED-MATRIX-CHAIN( $p$ )

```

1   $n = p.length - 1$ 
2  let  $m[1..n, 1..n]$  be a new table
3  for  $i = 1$  to  $n$ 
4    for  $j = i$  to  $n$ 
5       $m[i, j] = \infty$ 
6  return LOOKUP-CHAIN( $m, p, 1, n$ )
    
```

```

LOOKUP-CHAIN( $m, p, i, j$ )
1  if  $m[i, j] < \infty$ 
2    return  $m[i, j]$ 
3  if  $i == j$ 
4     $m[i, j] = 0$ 
5  else for  $k = i$  to  $j - 1$ 
6     $q =$  LOOKUP-CHAIN( $m, p, i, k$ )
7      + LOOKUP-CHAIN( $m, p, k + 1, j$ ) +  $p_{i-1} p_k p_j$ 
8    if  $q < m[i, j]$ 
9       $m[i, j] = q$ 
    return  $m[i, j]$ 
    
```

The MEMOIZED-MATRIX-CHAIN procedure, like MATRIX-CHAIN-ORDER, maintains a table  $m[1..n, 1..n]$  of computed values of  $m[i, j]$ , the minimum number of scalar multiplications needed to compute the matrix  $A_{i..j}$ . Each table entry initially contains the value  $\infty$  to indicate that the entry has yet to be filled in. Upon calling LOOKUP-CHAIN( $m, p, i, j$ ), if line 1 finds that  $m[i, j] < \infty$ , then the procedure simply returns the previously computed cost  $m[i, j]$  in line 2. Otherwise, the cost is computed as in RECURSIVE-MATRIX-CHAIN, stored in  $m[i, j]$ , and returned. Thus, LOOKUP-CHAIN( $m, p, i, j$ ) always returns the value of  $m[i, j]$ , but it computes it only upon the first call of LOOKUP-CHAIN with these specific values of  $i$  and  $j$ .

Figure 15.7 illustrates how MEMOIZED-MATRIX-CHAIN saves time compared with RECURSIVE-MATRIX-CHAIN. Shaded subtrees represent values that it looks up rather than recomputes.

Like the bottom-up dynamic-programming algorithm MATRIX-CHAIN-ORDER, the procedure MEMOIZED-MATRIX-CHAIN runs in  $O(n^3)$  time. Line 5 of MEMOIZED-MATRIX-CHAIN executes  $\Theta(n^2)$  times. We can categorize the calls of LOOKUP-CHAIN into two types:

- calls in which  $m[i, j] = \infty$ , so that lines 3–9 execute, and
- calls in which  $m[i, j] < \infty$ , so that LOOKUP-CHAIN simply returns in line 2.

Dependency  
(order subpb comp)

↔ recurse stack

if  $m[i, j]$  already computed  
not make recursive call

Memoization

$C =$  still a table

$C[i, j]$  cache computed values

never compute again  $C$

never recurse on computed value



only recurse on non-computed

$C[i, j]$  call

THAT ARE NEEDED

Memorization speeds up when it does not solve all subPBs in dependency table

PB

Memorization **Must solve** all subPBs anyway?

Exercise: fill this table

Rod Cut

Coin Change

Check Board

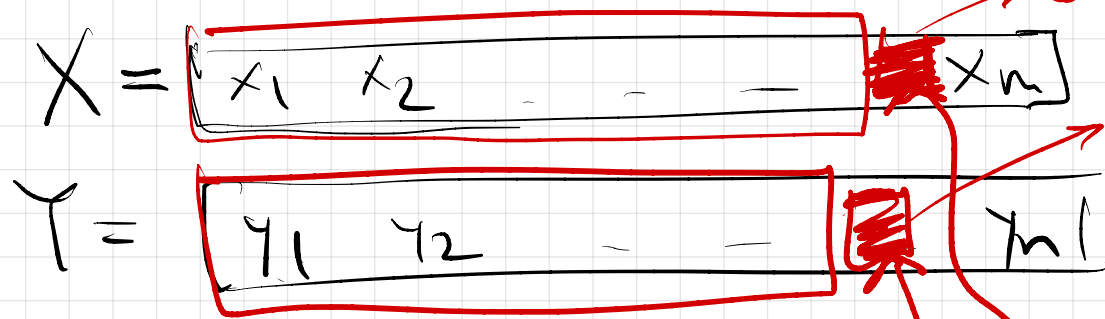
0/1 Knapsack

Matrix Chain

LCS

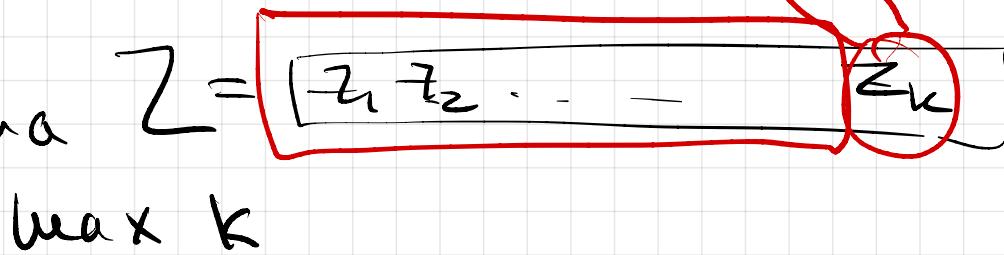
Optimal BST

LCS = longest common subsequence



$x_i = X[1:i]$  prefix based  
 $y_j = Y[1:j]$  given index  
 last occ( $z_k$ )

Common subsequence



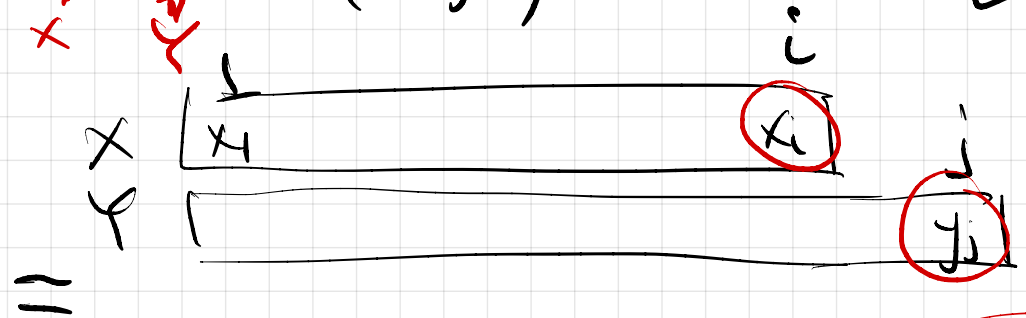
$Z$  subseq of  $X$   
 $Z$  subseq of  $Y$

① OPT SOL structure =  $Z \Rightarrow Z_{k+1} = Z[1:k]$  opt solution  
 to some  $X[1:?]$ ,  $Y[1:?]$   
 prefixes up to  $z_k$  value

assume

$z_k$  occurs in  $X$  and  $Y$

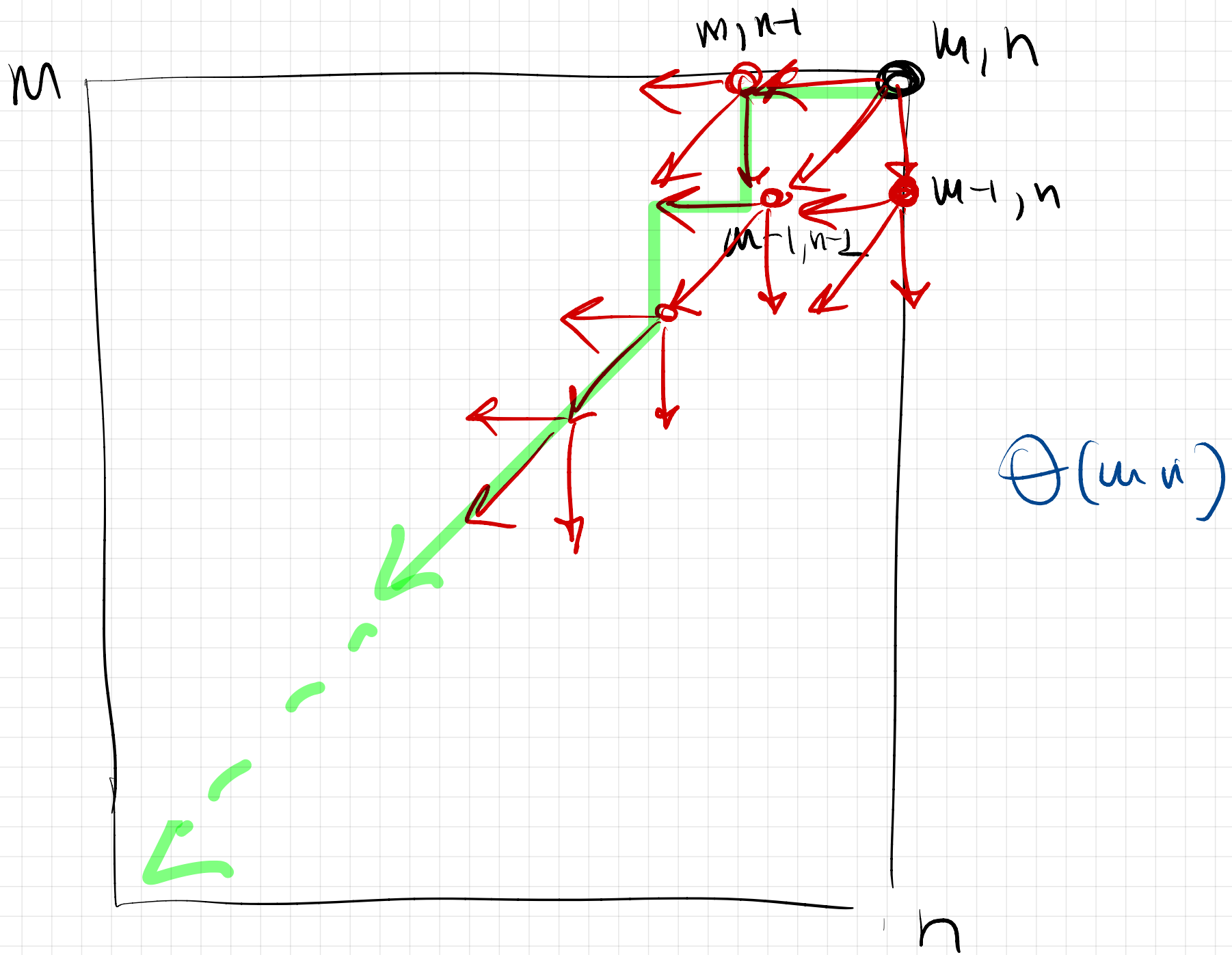
②A  $C[i][j] = \text{max of common subseq} \left\{ \begin{matrix} X[1:i] \\ Y[1:j] \end{matrix} \right\}$   
 (length) LCS



$z_k$  is last found  $\iff x_i = y_j$   $C[i-1, j-1] + 1$   
 $x_i \neq y_j$   $\begin{cases} \text{cut } x_i \\ \text{cut } y_j \end{cases} \max \left\{ \begin{matrix} C[i-1, j] \\ C[i, j-1] \end{matrix} \right\} \theta(1)$

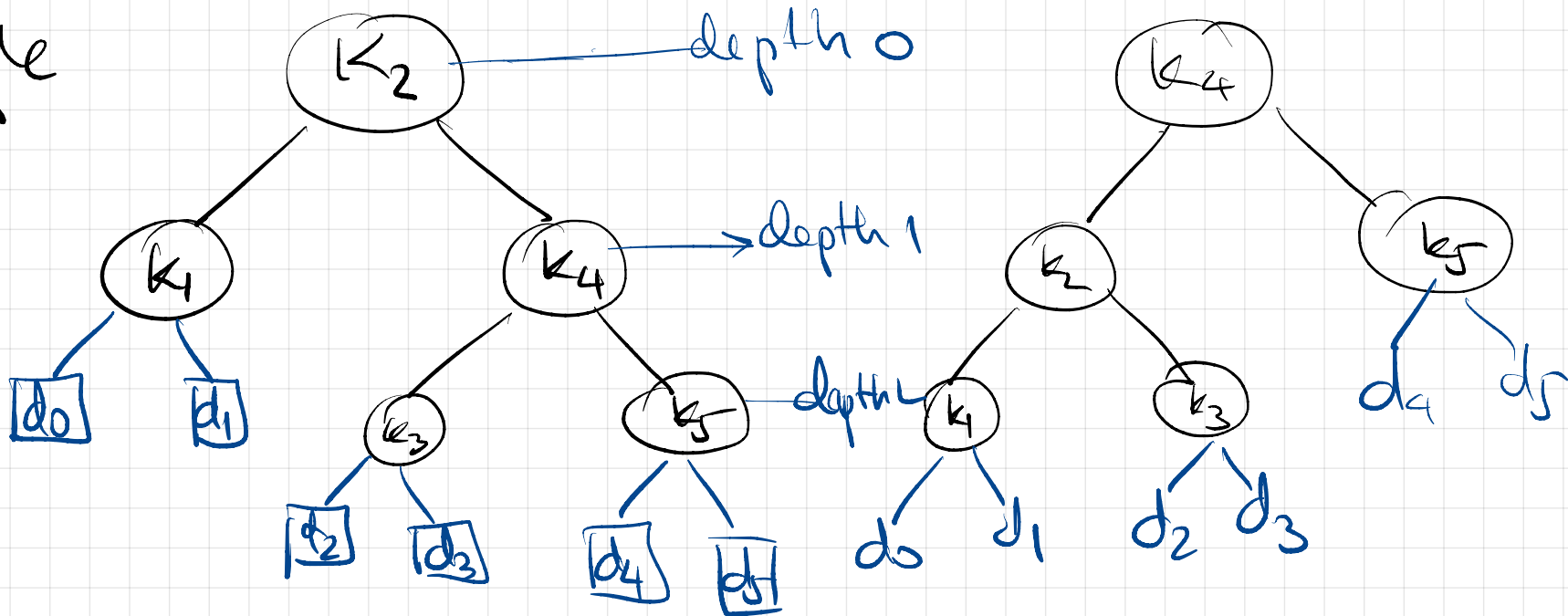
②B

$S[i][j] =$  which  $\text{max}$  is  $\text{max-obj}$   
 $\left\{ \begin{matrix} a \leftarrow a \leftarrow a \leftarrow a \leftarrow a \\ \downarrow \end{matrix} \right.$



Optimal BST ordered values  $k_1 \leq k_2 \leq \dots \leq$

example  
BST



Search probability:  $\Pr(k_i) = p_i$  *not uniform*  
 $\Pr(d_i) = q_i$   $d_i =$  searches for values  $k_i < \text{val} < k_{i+1}$

$$\sum p_i + \sum q_i = 1$$

depth = level

OPTIMALITY: min expected search cost

$$\sum_{i=1}^n [\text{depth}(k_i) + 1] p_i + \sum_{i=0}^n [\text{depth}(d_i) + 1] q_i$$

Step 1 OPT SOL given

$k_1 < \dots <$   
LEFT SUBSPS

$k_r < \dots < k_n$   
Right subsp.

$p_{bl}$   
 $0.6$

$0.4$   $p_{br}$

$k[1:r-1]$   
 $1 \leq k \leq r-1$   
 $d[0:r-1]$

$k[r+1:n]$   
 $d[r:n]$

$p_1 \dots p_{r-1}$   
 $q_0 \dots q_{r-1}$  } example  
 $0.6 \approx p_{bl}$

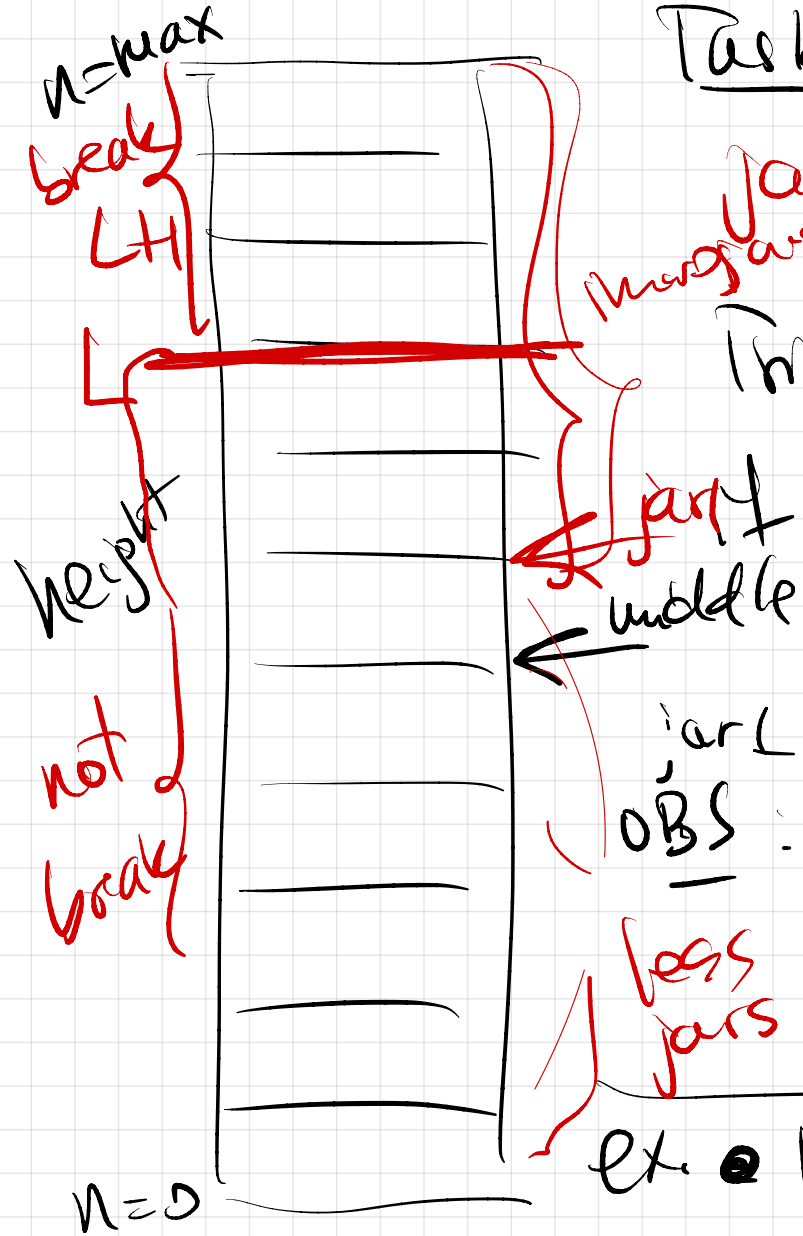
$p_{br}$   
 $0.4$  }  $p_{r+1} \dots p_n$   
 $1-p_{bl}$  }  $q_r \dots q_n$

Search:  $0.6 \cdot \downarrow_{root}$   
best way to search ( $k[1:r-1]$ )

Search:  $0.4 \cdot \downarrow_{root}$   
best way to search ( $k[r+1:n]$ )

days on ladder  $n$  steps  $k$  jars

Task : find highest level  $L$  where jars don't break (they break at  $L+1$ )



That jar at level  $l$

$l \leq L \Rightarrow$  doesn't break  $\Rightarrow$  can reuse it

$l > L+1 \Rightarrow$  jar breaks cannot reuse it

OBS : minimum # trials in worst case

You must 100% Find  $L$

ex. •  $k=1 \Rightarrow$  bottom up on stair  $\Theta(n)$

•  $k > \log n \Rightarrow$  binary search