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Introduction to

Algorithm Design and Analysis

[17] Dynamic Programming 2



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In the Last Class...

- **Basic idea of DP**
- **Least cost matrix multiplication**
 - BF1, BF2
 - A DP solution
- **Weighted binary search tree**
 - The same DP solution



DP - II

- All-pair shortest path
- Edit distance
- Highway restaurants
- Separating sequence of words
- Changing coins
- Elements of DP



All-pair Shortest Path

- **“BF 2”**
 - Path length k
 - k in $[1, n]$
- **Floyd algorithm**
 - Index range k
 - k in $[1, n]$



BF2

$$dist(u, v, k) = \begin{cases} 0 & \text{if } u = v \\ \infty & \text{if } k = 0 \text{ and } u \neq v \\ \min_x (dist(u, x, k-1) + w(x \rightarrow v)) & \text{otherwise} \end{cases}$$

APSP(V, E, w):

```

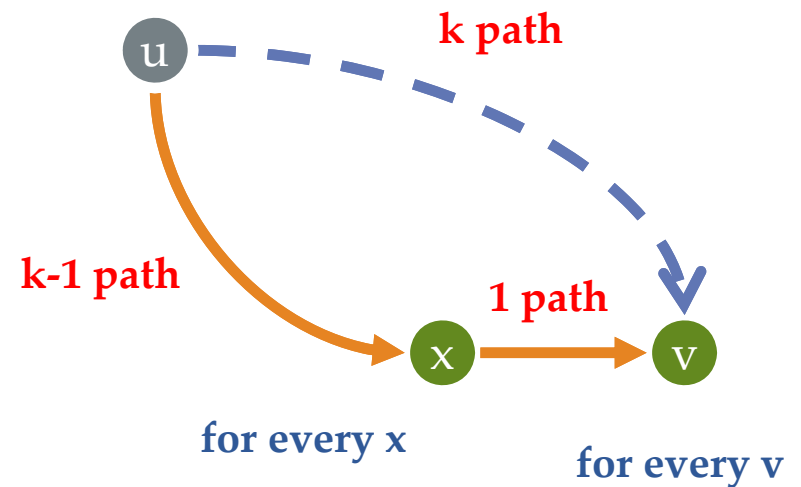
for all vertices u
  for all vertices v
    if u = v
      dist[u, v, 0] ← 0
    else
      dist[u, v, 0] ← ∞
  for k ← 1 to V - 1
    for all vertices u
      for all vertices v
        for all vertices x
          if dist[u, v, k] > dist[u, x, k-1] + w(x → v)
            dist[u, v, k] ← dist[u, x, k-1] + w(x → v)

```

$O(n^4)$

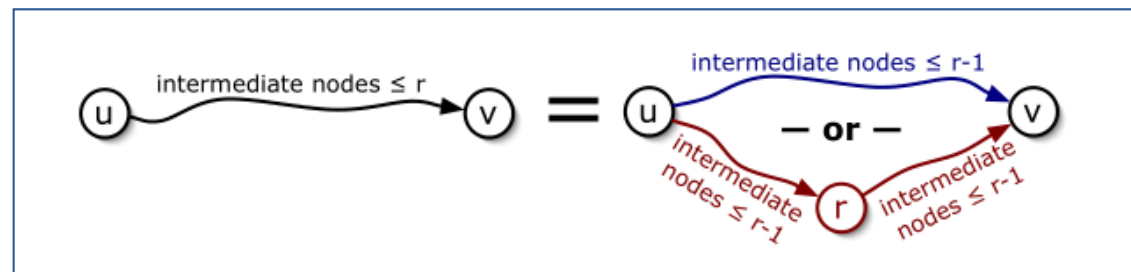
Length of the shortest
of at most k edges

for every u



Floyd Algorithm

- Basic idea



- Smart recursion

$$dist(u, v, r) = \begin{cases} w(u \rightarrow v) & \text{if } r = 0 \\ \min \{ dist(u, v, r-1), dist(u, r, r-1) + dist(r, v, r-1) \} & \text{otherwise} \end{cases}$$



Floyd Algorithm

- Basic DP (3-dimensional)

```
FLOYDWARSHALL( $V, E, w$ ):  
  for all vertices  $u$   
    for all vertices  $v$   
       $dist[u, v, 0] \leftarrow w(u \rightarrow v)$   
  
  for  $r \leftarrow 1$  to  $V$   
    for all vertices  $u$   
      for all vertices  $v$   
        if  $dist[u, v, r - 1] < dist[u, r, r - 1] + dist[r, v, r - 1]$   
           $dist[u, v, r] \leftarrow dist[u, v, r - 1]$   
        else  
           $dist[u, v, r] \leftarrow dist[u, r, r - 1] + dist[r, v, r - 1]$ 
```

$O(n^3)$

- Improved DP (2-dimensional)

```
FLOYDWARSHALL2( $V, E, w$ ):  
  for all vertices  $u$   
    for all vertices  $v$   
       $dist[u, v] \leftarrow w(u \rightarrow v)$   
  
  for all vertices  $r$   
    for all vertices  $u$   
      for all vertices  $v$   
        if  $dist[u, v] > dist[u, r] + dist[r, v]$   
           $dist[u, v] \leftarrow dist[u, r] + dist[r, v]$ 
```

$O(n^3)$



Edit Distance

- You can edit a word by
 - Insert, Delete, Replace
- Edit distance
 - Minimum number of edit operations
- Problem
 - Given two strings, compute the edit distance

F	O	O		D
M	O	N	E	Y

The edit distance is 4

4 op: **R** **R** **I** **R**

3 op: not possible

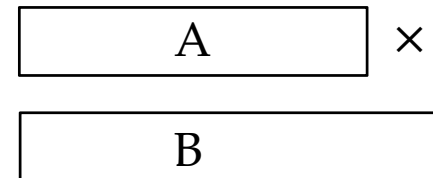


“BF” Recursion

- **Case 1**

- 1.1 Insert
- 1.2: dual of case 1.1

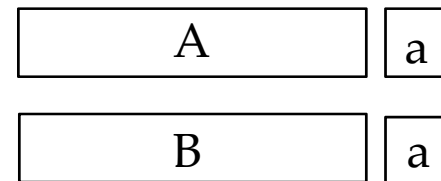
Case 1.1



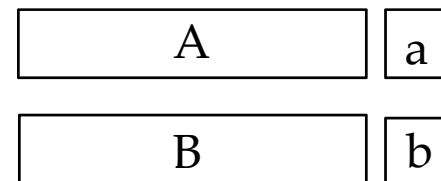
- **Case 2**

- 2.1 $a=a$
- 2.2 $a \neq b$

Case 2.1



Case 2.2



“BF” Recursion

- **EditDis(i,j)**
 - Base case:
 - If $i=0$, $\text{EditDis}(i,j)=j$
 - If $j=0$, $\text{EditDis}(i,j)=i$
 - Recursion:

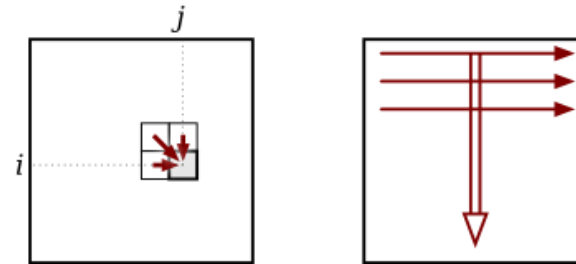
$$\text{EditDis}(A[1..m], B[1..n]) = \min \begin{cases} \text{EditDis}(A[1..m-1], B[1..n]) + 1 \\ \text{EditDis}(A[1..m], B[1..n-1]) + 1 \\ \text{EditDis}(A[1..m-1], B[1..n-1]) + I\{A[m] \neq B[n]\} \end{cases}$$



Smart Programming

- DP dict
 - EditDis[1..m, 1..n]
- DP algorithm

dependencies



```
EDITDISTANCE(A[1..m], B[1..n]):  
  for j ← 1 to n  
    Edit[0, j] ← j  
  for i ← 1 to m  
    Edit[i, 0] ← i  
    for j ← 1 to n  
      if A[i] = B[j]  
        Edit[i, j] ← min {Edit[i - 1, j] + 1, Edit[i, j - 1] + 1, Edit[i - 1, j - 1]}  
      else  
        Edit[i, j] ← min {Edit[i - 1, j] + 1, Edit[i, j - 1] + 1, Edit[i - 1, j - 1] + 1}  
  return Edit[m, n]
```

Example

algorithm

vs.

altruistic

		A	L	G	O	R	I	T	H	M
	0	→ 1	→ 2	→ 3	→ 4	→ 5	→ 6	→ 7	→ 8	→ 9
A	1	↓ 0	→ 1	→ 2	→ 3	→ 4	→ 5	→ 6	→ 7	→ 8
L	2	↓ 1	↓ 0	→ 1	→ 2	→ 3	→ 4	→ 5	→ 6	→ 7
T	3	↓ 2	↓ 1	↓ 1	→ 2	→ 3	→ 4	→ 4	→ 5	→ 6
R	4	↓ 3	↓ 2	↓ 2	↓ 2	→ 3	→ 4	→ 5	→ 6	
U	5	↓ 4	↓ 3	↓ 3	↓ 3	↓ 3	→ 4	→ 5	→ 6	
I	6	↓ 5	↓ 4	↓ 4	↓ 4	↓ 4	↓ 3	→ 4	→ 5	→ 6
S	7	↓ 6	↓ 5	↓ 5	↓ 5	↓ 5	↓ 4	↓ 4	→ 5	→ 6
T	8	↓ 7	↓ 6	↓ 6	↓ 6	↓ 6	↓ 5	↓ 4	→ 5	→ 6
I	9	↓ 8	↓ 7	↓ 7	↓ 7	↓ 7	↓ 6	↓ 5	↓ 5	→ 6
C	10	↓ 9	↓ 8	↓ 8	↓ 8	↓ 8	↓ 7	↓ 6	↓ 6	↓ 6



DP in One Dimension

- **Highway restaurants**
 - n possible locations on a straight line
 - $m_1, m_2, m_3, \dots, m_n$
 - At most one restaurant at one location
 - Expected profit for location i is p_i
 - Any two restaurants should be at least k miles apart
- **How to arrange the restaurants**
 - To obtain the maximum expected profit



Highway Restaurants

- **The recursion**
 - $P(j)$: the max profit achievable using only first j locations
 - $P(0)=0$
 - $\text{prev}[j]$: largest index before j and k miles away

$$P(j) = \max(p_j + P(\text{prev}[j]), P(j - 1))$$



Highway Restaurants

- One dimension DP algorithm
 - Fill in $P[0], P[1], \dots, P[n]$

```
(First compute the prev[.] array)
i = 0
for j = 1 to n:
    while mi+1 ≤ mj - k:
        i = i + 1
    prev[j] = i
(Now the dynamic programming begins)
P[0] = 0
for j = 1 to n:
    P[j] = max(pj + P[prev[j]], P[j - 1])
return P[n]
```



Words into Lines

- **Words into lines**
 - Word-length w_1, w_2, \dots, w_n and line-width: W
- **Basic constraint**
 - If w_i, w_{i+1}, \dots, w_j are in one line, then $w_i + w_{i+1} + \dots + w_j \leq W$
- **Penalty for one line: some function of X . X is:**
 - 0 for the last line in a paragraph, and
 - $W - (w_i + w_{i+1} + \dots + w_j)$ for other lines
- **The problem**
 - How to make the penalty of the paragraph, which is the sum of the penalties of individual lines, minimized



Greedy Solution

i	word	w
1	Those	6
2	who	4
3	cannot	7
4	remember	9
5	the	4
6	past	5
7	are	4
8	condemned	10
9	to	3
10	repeat	7
11	it.	4

Solution by greedy strategy

words	(1,2,3)	(4,5)	(6,7)	(8,9)	(10,11)
X	0	4	8	4	0
penalty	0	64	512	64	0

Total penalty is **640**

An improved solution

words	(1,2)	(3,4)	(5,6,7)	(8,9)	(10,11)
X	7	1	4	4	0
penalty	343	1	64	64	0

Total penalty is **472**

W is 17, and penalty is X^3



Problem Decomposition

- Representation of subproblem: a pair of indexes (i,j) , breaking words i through j into lines with minimum penalty.
- Two kinds of subproblem
 - (k, n) : the penalty of the last line is 0
 - all other subproblems
- For some k , the combination of the optimal solution for $(1,k)$ and $(k+1,n)$ gives a optimal solution for $(1,n)$.
- Subproblem graph
 - About n^2 vertices
 - Each vertex (i,j) has a edge to about $j-i$ other vertices, so, the number of edges is in $\Theta(n^3)$



Simpler Identification of Subproblems

- If a subproblem concludes the paragraph, then (k,n) can be simplified as (k)
 - About k subproblems
- Can we eliminate the use of (i,j) with $j < n$?
 - Put the first k words in the first line (with the basic constraint satisfied), the subproblem to be solved is $(k+1,n)$
 - Optimizing the solution over all k 's. (k is at most $W/2$)



Breaking into Lines

```
lineBreak( $w, W, i, n, L$ )  
  if ( $w_i + w_{i+1} + \dots + w_n \leq W$ )  
    <Put all words on line  $L$ , set penalty to 0>  
  else  
    for ( $k=1; w_i + \dots + w_{i+k-1} \leq W; k++$ )  
       $X = W - (w_i + \dots + w_{i+k-1});$   
       $kPenalty = lineCost(X) + lineBreak(w, W, i+k, n, L+1)$   
      <Set penalty always to the minimum  $k$  Penalty>  
      <Updating  $k_{min}$ , which records the  $k$  that produced  
        the minimum penalty>  
      <Put words  $i$  through  $i+k_{min}-1$  on line  $L$ >  
    return penalty
```

In *DP* version this
is replaced by
“**Recursion or
Retrieve**”

In *DP*
version,
“**Storing**”
inserted



Converted to DP

- Subproblem Dictionary: $P[1], P[2], \dots, P[n]$
- lineBreakDP

```
for(i=n; i>=1;i--){
    if (all words in one line){
        P[i]=0;
        continue;
    } // fi
    while (sum of word length <= W){
        // assume we have k words put in the current line
        calculate the penalty  $\text{Cost}_{\text{cur}}$  of putting k words in this line;
         $\text{minCost} = \min\{\text{minCost}, \text{Cost}_{\text{cur}} + P[i+k]\}$  if necessary;
    } // while
    P[i]=minCost;
} // for
```



Analysis of lineBreak()

- Each subproblem is identified by only one integer k , for (k, n)
 - Number of vertex in the subproblem graph: at most n
 - So, in **DP** version, the recursion is executed at most n times.
- So, the running time is in $\Theta(Wn)$
 - The loop is executed at most $W/2$ times.
 - In fact, W , the line width, is usually a constant. So, **$\Theta(n)$** .
 - The extra space for the dictionary is in **$\Theta(n)$** .



Making Change: Revisited

- **How to pay a given amount of money?**
 - Using the smallest possible number of coins
 - With certain systems of coinage
- **We have known that the greedy strategy fails sometimes**



Subproblems

- **Assumptions**

- Given n different denominations
- A coin of denomination i has d_i units
- The amount to be paid: N .

- **Subproblem $[i,j]$**

- The minimum number of coins required to pay an amount of j units, using only coins of denominations 1 to i .

- **The problem**

- Figure out subproblem $[n, N]$ (as $c[n,N]$)



Dependency of Subproblems

- $c[i,0]$ is 0 for all i
- When we are to pay an amount j using coins of denominations 1 to i , we have two choices:
 - No coins of denomination i is used: $c[i-1, j]$
 - One coins of denomination i is used: $1+c[i, j-d_i]$
- So, $c[i,j] = \min (c[i-1, j], 1+c[i, j-d_i])$



Data Structure

Define a array $\text{coin}[1..n, 0..N]$ for all $c[i, j]$

an example

	0	1	2	3	4	5	6	7	8
$d_1=1$	0	1	2	3	4	5	6	7	8
$d_2=4$	0	1	2	3	1	2	3	4	2
$d_3=6$	0	1	2	3	1	2	1	2	2

direction of computation



The Procedure

```
int coinChange(int N, int n, int[] coin)
    int denomination[]=[ $d_1, d_2, \dots, d_n$ ];
    for ( $i=1; i \leq n; i++$ )
        coin[ $i,0$ ]=0;
        for ( $i=1; i \leq n; i++$ )
            for ( $j=1; j \leq N; j++$ )
                if ( $i=1 \ \&\& \ j < \text{denomination}[i]$ ) coin[ $i,j$ ]= $+\infty$  ;
                else if ( $i=1$ ) coin[ $i,j$ ]=1+coin[1, $j-\text{denomination}[1]$ ];
                else if ( $j < \text{denomination}[i]$ ) coin[ $i,j$ ]=cost[ $i-1, j$ ];
                else coin[ $i,j$ ]=min(coin[ $i-1, j$ ], 1+coin[ $i, j-\text{denomination}[i]$ ];
    return coin[ $n,N$ ];
```

in $\Theta(nM)$,
 n is usually a constant



Other DP Problems

- **Text string problems**
 - Longest common subsequence, ...
 - Variations of standard text string problems
- **One dimensional problems**
 - Arrangements along the road, ...
- **Graph problems**
 - Shortest paths over DAGs, ...
- **Hard problems**
 - Knapsack problems and variations, ...



Principle of Optimality

- Given an optimal sequence of decisions, each *subsequence must be optimal by itself*.

- Pos

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- some of its sub-instances.

- It is often not obvious which sub-instances are relevant to the instance under consideration.

Optimal
Substructure

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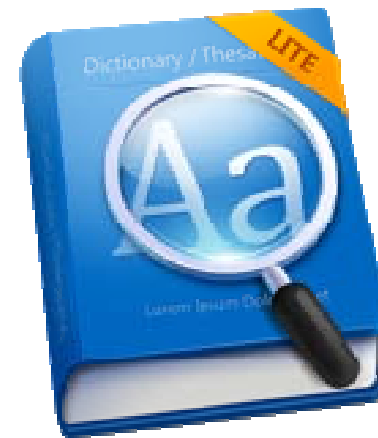
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Elements of Dynamic Programming

- Symptoms of DP
 - Overlapping subproblems
 - Optimal substructure
- How to use DP
 - “**Brute force**” recursion
 - Overlapping subproblems
 - **Smart** programming
 - Topological ordering of subproblems

DP Dictionary



Thank you!

Q & A

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