

DYNAMIC PROGRAMMING

1) Longest path in a DAG $G = (V, E)$

Subproblem: $L(v) = \text{length of longest path ending in } v$

Recurrency: $L(v) = \max_{(u, v) \in E} (L(u) + 1)$ (where an empty max equals 0)

Order to compute: topological sorted order of G

Run Time: $O(|V| + |E|)$

2) Longest Increasing Subsequence of sequence a_1, \dots, a_n

Subproblem: $L(i) = \text{length of longest increasing subsequence ending in } a_i$

Recurrency: $L(i) = 1 + \max_{j < i} L(j) \mathbb{1}_{a_j < a_i}$

Order to compute: $i = 1, 2, 3, \dots$

Run Time: $O(n^2)$

3) Edit Distance of two strings $x[1:n], y[1:m]$

$E(x[1:n], y[1:m]) = \min \left\{ \begin{array}{l} \# \text{ of deletes, inserts, substitutes} \\ \text{to transform } x[1:n] \text{ into } y[1:m] \end{array} \right.$

Subproblem: $E(x[1:i], y[1:j])$

Recurrency: $E(x[1:i], y[1:j]) = \min \left\{ \begin{array}{l} E(x[1:i], y[1:j-1]) + 1 \\ E(x[1:i-1], y[1:j]) + 1 \\ E(x[1:i-1], y[1:j-1]) + \mathbb{1}_{x_i \neq y_j} \end{array} \right.$

Order to compute:

	\emptyset	S	N	O	W	Y
\emptyset	0	1	2	3	4	5
S	1					
U	2					
N	3					
N	4					
Y	5					

	\emptyset	S	N	O	W	Y
\emptyset	0	1	2	3	4	5
S	1					
U	2					
N	3					
N	4					
Y	5					

	\emptyset	S	N	O	W	Y
\emptyset	0	1	2	3	4	5
S	1					
U	2					
N	3					
N	4					
Y	5					

RunTime $O(nm)$

$O(nm)$

$O(n+m)$ in parallel

space $O(m)$

$O(n)$

$O(n+m)$

4) Knapsack

Input: Capacity W (integer)

weights w_1, \dots, w_n (integers)

values v_1, \dots, v_n (")

Goal: Find set of item with

max total value, with total weight $\leq W$

4a) Knapsack with Replacement

We looked optimal solution

$$v_{i_1} + \dots + v_{i_k} = v_{i_1} + \dots + v_{i_{k-1}} + v_{i_k} = K(W - w_{i_k}) + v_{i_k}$$

Subproblem:

$K(C) = \text{max total value with total weight } \leq C$

$$C = 0, 1, \dots, W$$

Recurrence:

$$K(C) = \max_{i: w_i \leq C} (v_i + K(C - w_i))$$

Order to compute

$C = 0, 1, \dots$, starting with $K(0) = 0$

Algorithm:

Input: $w, v[1:n], w[1:n]$

$$K(0) = 0$$

For $C = 1 \text{ to } W$

$$K(C) = \max_{i: w_i \leq C} v_i + K(C - w_i)$$

Output: $K(W)$

Runtime $O(nW)$ Space $O(W+n)$
 exponential in $\log W$

4b) Knapsack w/o replacement

1) Subproblem

Optimal solution: items i_1, \dots, i_n , $\sum w_{i_k} \leq W$

$$\text{Value} = v_{i_1} + \dots + v_{i_n} = \underbrace{v_{i_1} + \dots + v_{i_{n-1}}}_{\tilde{K}(W-w_{i_n})} + v_{i_n}$$

Guess: $\tilde{K}(W-w_{i_n})$

$\tilde{K}(C)$ optimum w/o replacement at capacity C

Problem: i_1, \dots, i_{n-1} must be different from i_n , but in $\tilde{K}(W-w_{i_n})$ i_n could be used again

2nd Try:

$$\tilde{K}(k) = \text{Knapsack}(v_1, \dots, v_k, w_1, \dots, w_k)$$

$$\tilde{K}(n) = \max \{ \tilde{K}(n-1), v_n + \tilde{K}(n-1) \}$$

Problem: We are not keeping track of the budget

3rd Try:

$$K(C, k) = \text{Optimum with total weight } \leq C$$

using a subset of items $1, \dots, k$

2) Recursion

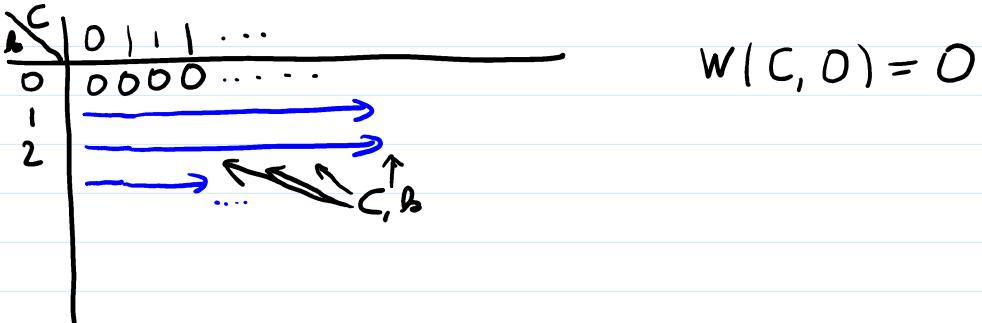
$$(K(C, k-1))$$

if $v_k > C$

2) Recursion

$$K(C, k) = \begin{cases} K(C, k-1) & \text{if } v_k > C \\ \max\{K(C, k-1), K(C - w_k, k-1) + v_k\} & \text{if } v_k \leq C \end{cases}$$

3) Order



Algorithm:

Input: $W, v[1:n], w[1:n]$

For $C = 0, 1, \dots, W$

$$K(C, 0) = 0$$

For $k = 1, \dots, n$ "Calculate $W(k, \cdot)$ "

For $C = 1, \dots, W$

$$K(C, k) = K(C, k-1)$$

IF $w_k \leq C$ and $v_k + K(C - w_k, k-1) > K(C, k-1)$

$$K(C, k) = v_k + K(C - w_k, k-1)$$

Output $K(W, n)$

Runtime
 $O(nW)$

Space

~~$O(nW)$~~ $O(n+W)$
better implementation

5) Single Source Shortest Path

Input: Weighted graph $G = (V, E, \ell)$, source s (where $\ell(u, v) \in \mathbb{R}$)

Output: $\forall v \in V$, $\text{dist}(v) = \text{length of shortest path } s \rightsquigarrow v$

1) Subproblems

Optimal solution:

$$v_1 = s, v_2, \dots, v_k = v$$

Optimal Solution:

$$v_1 = s, v_2, \dots, v_n = v$$

$$\text{dist}(v) = \text{dist}(v_{n-1}) + \ell(v_{n-1}, v)$$

$$\Rightarrow \text{dist}(v) = \min_u (\text{dist}(u) + w_{uv}) \quad \text{not really a subproblem!}$$

Idea: recurse on # of edges k in path

$\text{dist}(v, k) = \text{length of shortest path } s \rightsquigarrow v \text{ using } k \text{ edges}$

$$k=0, 1, \dots$$

Recursion

$$\text{dist}(v, k) = \min \left\{ \text{dist}(v, k-1), \min_{u \in E} (\text{dist}(u, k-1) + \ell(u, v)) \right\}$$

Dependencies

$\text{dist}(v, k)$ depends on $\text{dist}(u, k-1)$ for $u \in E$

Algorithm

For $v \in V$: $\text{dist}(v, 0) = \infty$

$$\text{dist}(v, 0) = 0$$

$$\text{For } k=1, \dots, n-1$$

For all $v \in V$

$$\text{dist}(v, k) = \text{dist}(v, k-1)$$

For all $(u, v) \in E$

$$\text{IF } \text{dist}(v, k) > \text{dist}(u, k-1) + \ell(u, v)$$

$$\text{dist}(v, k) = \text{dist}(u, k-1) + \ell(u, v)$$

$\forall v \in V$ Output $\text{dist}(v, n-1)$

Runtime

$$O(|V|(|E| + |V|)) \approx O(|E||V|)$$

Rem: This is very similar to Bellman Ford, in fact, in the worst case, it is identical. (sometimes Bellman Ford just needs one run, e.g. if $G = \text{---} \circ \text{---}$, in which case B.F. will be done after one run through all edges, while D.P. needs $n-1$ runs)

6) All Pairs Shortest path

Input: $G = (V, E)$ $\ell: E \rightarrow \mathbb{R}$

Output: $\text{dist}(u, v) = \text{length of shortest path from } u \text{ to } v$
for all $u, v \in V$

Ideal: Run Bellman-Ford $|V|$ times Run Time $O(|V|^2 |E|)$

Today: DP Solution Floyd Warshall $\rightarrow O(|V|^3)$

Optimal Solution:

$$\text{dist}(u, v) = \ell(u, v_1) + \ell(v_1, v_2) + \dots + \ell(v_k, v)$$

Idea 1:

$$\text{dist}(u, v, k) = \underbrace{\# \text{ of edges used}}_{\# \text{ of intermediate vertices} + 1}$$

Problem: by iteration on the number of intermediate edges, we effectively ran the algorithm from 5 for all sources n , getting again $O(|V|^2 |E|)$ run-time.

Idea 2:

Iterate on which vertices are used

Subproblem

$$V = \{1, 2, \dots, n\}$$

$\text{dist}(i, j; k)$ uses intermediate vertices $\leq k$

Base Case

$$\text{dist}(i, j; 0) = \ell(i, j)$$

Recursion:

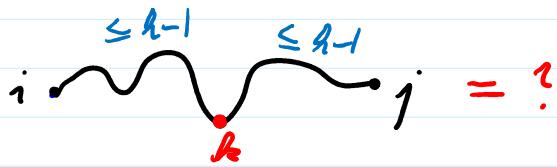
Assume $\text{dist}(i, j, k-1)$ is known

$$\text{dist}(i, j, k) = \min_{\substack{\text{path } i \rightarrow j \\ \text{with } \leq k \text{ edges}}} \dots$$

Case 1: k is not used

$$= \text{dist}(i, j, k-1)$$

Case 2: k is used



$$\text{dist}(i, j, k) = \min \{ \text{dist}(i, j, k-1), \text{dist}(i, k, k-1) + \text{dist}(k, j, k-1) \}$$

Algorithm:

FW(G, L)

For $i, j = 1, \dots, n$ $\text{dist}(i, j, 0) = \infty$

For $(i, j) \in E$ $\text{dist}(i, j, 0) = l(i, j)$

For $i = 1, \dots, n$ $\text{dist}(1, i, 0) = 0$

For $k = 1, \dots, n$

For $i = 1, \dots, n$

For $j = 1, \dots, n$

$$\text{dist}(i, j, k) = \min \{ \text{dist}(i, j, k-1), \text{dist}(i, k, k-1) + \text{dist}(k, j, k-1) \}$$

Output $\text{dist}(i, j, k)$

7) Travelling Salesman Problem (TSP)

Given: n cities, distances d_{ij} $i \neq j$

Goal: Find path of minimal length,

starting at 1, ending at 1, visiting every city once

Naive Running Time $n! = O((\frac{n}{e})^n)$

Goal: Find optimal tour in time $O(2^n)$

Idea: Consider subproblem on

set of cities $S \subseteq \{1, \dots, n\}$, $1 \in S$

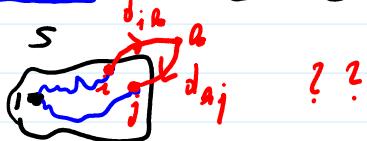
Ideal 1:

$r(S) = \text{cost of tour in } S \text{ starting}$

Idea 1:

$C(S)$ = cost of tour in S , starting
and ending in l

Problem: $S \rightarrow S \cup \{k\}$



Idea 2:

$C(S) = \min_{j \neq l} \text{Cost}(\text{Tour } l \rightarrow j \text{ in } S) + d_{j \rightarrow l}$

Subproblem:

Let $l, j \in S, j \neq l$

$C(S, j) = \text{Length of shortest path from } l \text{ to } j$
visiting every $i \in S$ once