

More Dynamic Programming

Common Subproblems

- $\text{Opt}(i)$ - Opt solution using $x_1, \dots, x_i$. (eg LIS, longest path).
- $\text{Opt}(i, j)$ - Opt solution using $x_i, \dots, x_j$. (eg RNA)
- $\text{Opt}(i, j)$ - Opt solution using $x_1, \dots, x_i$ and $y_1, \dots, y_j$. (eg Edit distance)
- $\text{Opt}(r)$ - Opt solution using subtree rooted at r . (eg Vertex cover on trees).

Longest increasing subsequence

Given: sequence of numbers

Goal: find longest increasing subsequence

41 , 22 , 9 , 15 , 23 , 39 , 21 , 56 , 24 , 34 , 59 , 23 , 60 , 39 , 87 , 23 , 90

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longest increasing subsequence: length 9

Longest increasing subsequence

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Subproblems: $l(j)$ - length of longest increasing subseq. ending at j .

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Observation: if longest inc. sub. ending at j is $x_{i_1}, x_{i_2}, \dots, x_i, x_j$ then $l(j) = l(i) + 1$

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Claim:
$$l(j) = \begin{cases} 1 & \text{if } x_i \geq x_j, \text{ for all } i < j \\ 1 + \max_{i: i < j, x_i < x_j} l(i) & \text{else} \end{cases}$$

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Algorithm:

```
for  $j=1, \dots, n$ 
    if  $x_i \geq x_j$ , for all  $i < j$ , set  $l(j) = 1$ 
    else, set  $l(j) = 1 + \max_{i: i < j, x_i < x_j} l(i)$ 
output  $\max_j l(j)$ 
```

Running time

$O(n^2)$

All pairs shortest path in directed graph with no negative cycles.

Given: directed graph, (possibly negative) edge weights

Goal: find shortest path between every two vertices

Bellman-Ford algorithm can do this in time $O(n^2m)$

All pairs shortest path in directed graph with weighted edges

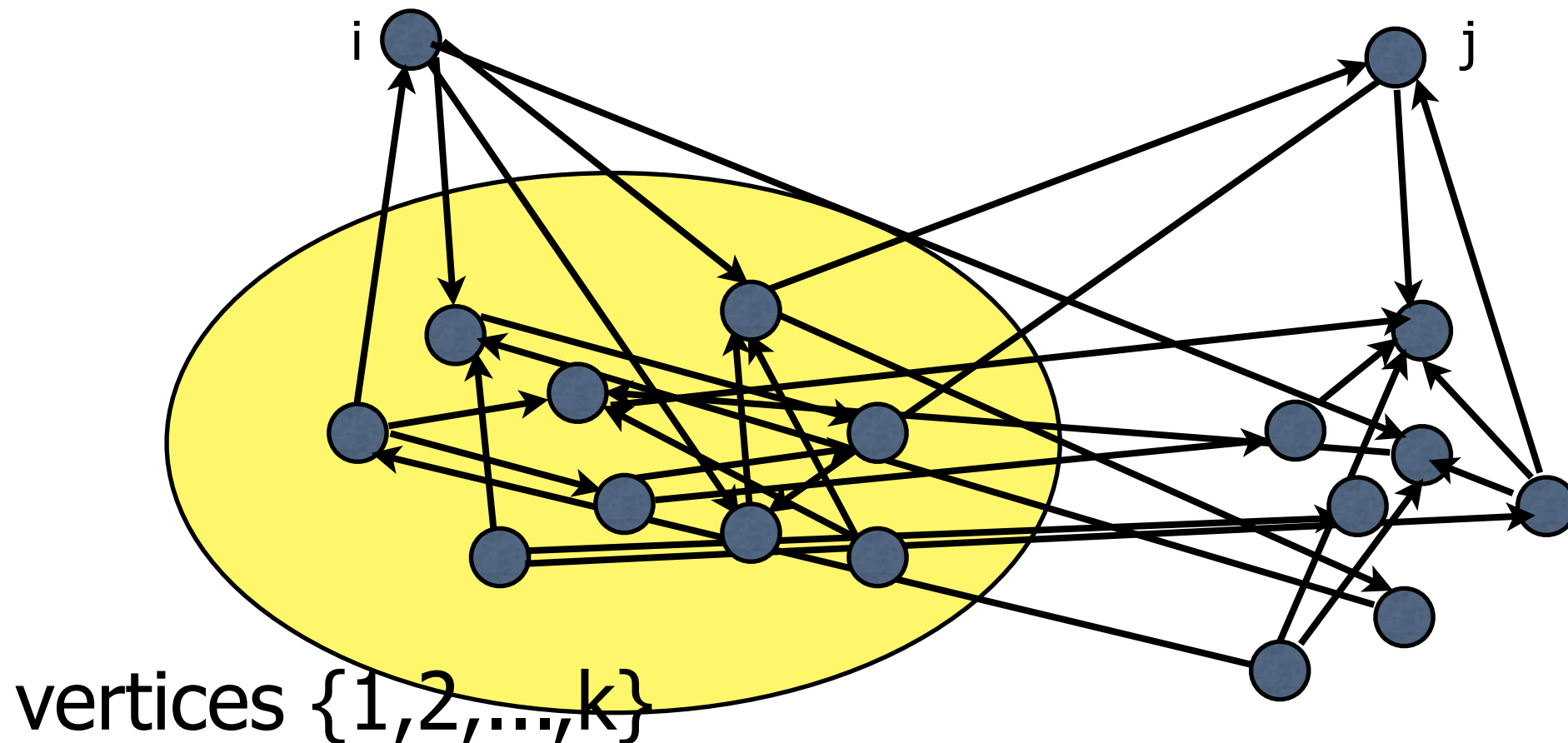
Given: directed graph, (possibly negative) edge weights

Goal: find shortest path between every two vertices

Subproblems: $d(i,j,k)$ - length of shortest path that starts at i , ends at j and visits only $\{1,2,\dots,k\}$ in the middle.

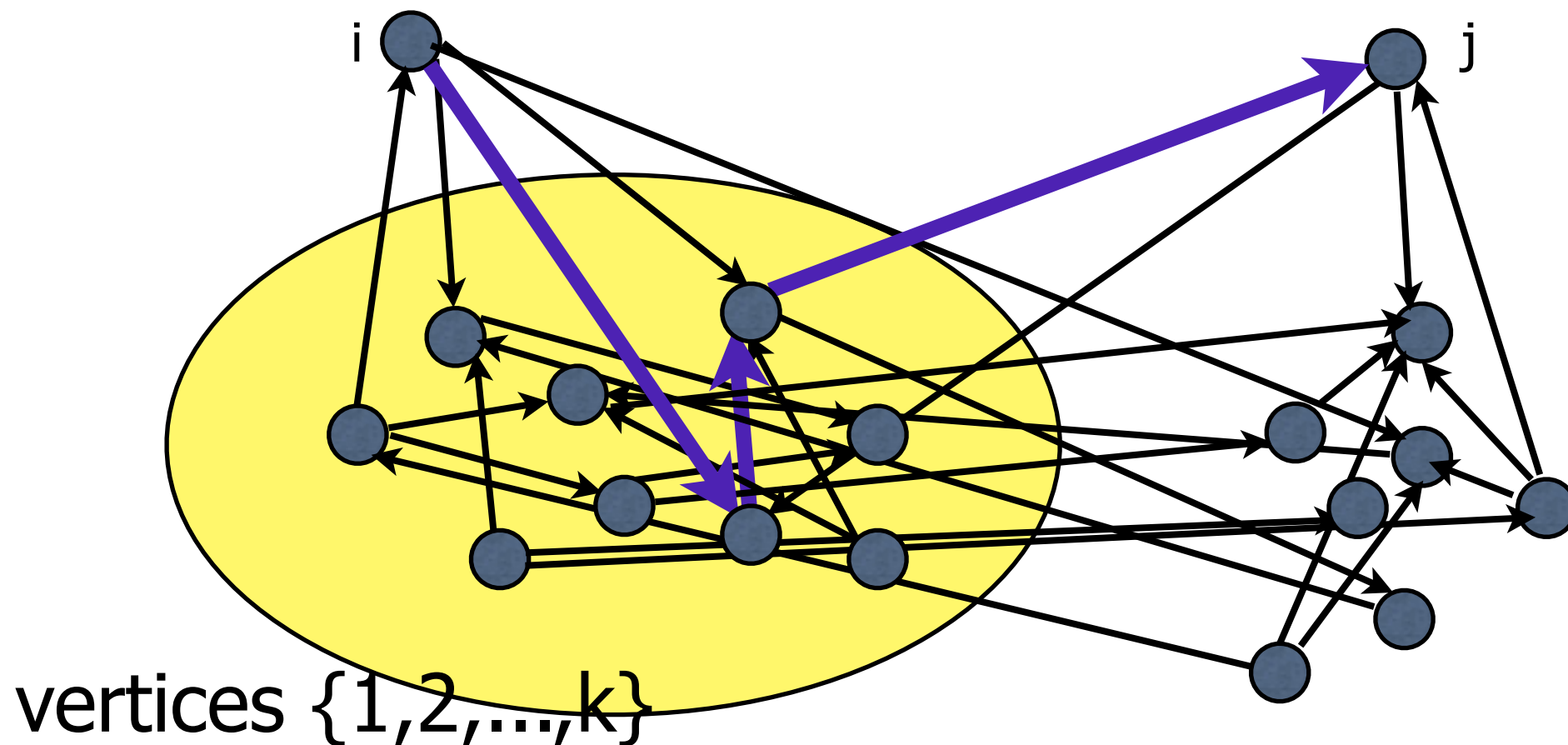
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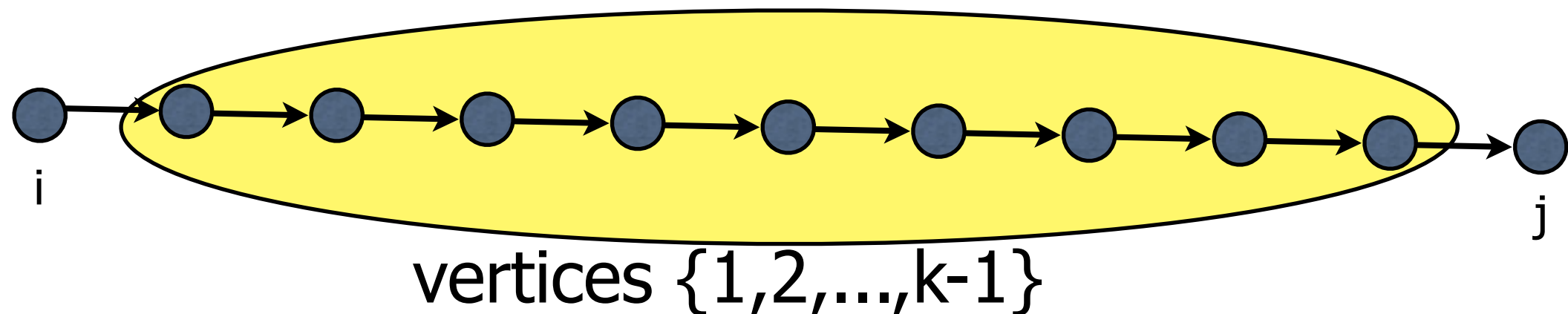
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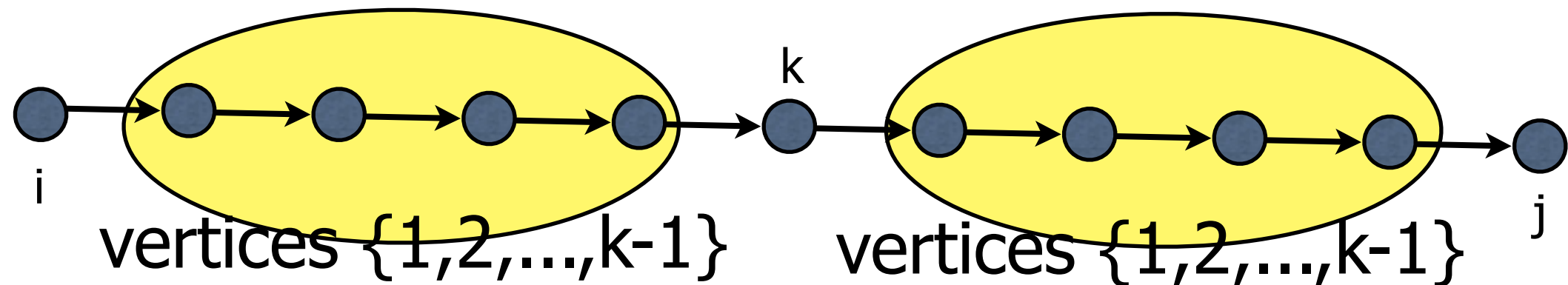
Observation:

if shortest path for $d(i,j,k)$ does not visit k , then
 $d(i,j,k) = d(i,j,k-1)$.



Otherwise,

$$d(i,j,k) = d(i,k,k-1) + d(k,j,k-1)$$



Subproblems: $d(i,j,k)$ - length of shortest path that starts at i , ends at j and every other vertex on path is in $\{1,2,\dots,k\}$.

Claim: $d(i,j,k) = \min\{d(i,j,k-1), d(i,k,k-1)+d(k,j,k-1)\}$

Algorithm:

```
for all  $i,j=1,\dots,n$ 
    set  $d(i,j,0)$  = weight of edge  $(i,j)$ 

for  $k=1,\dots,n$ 
    for all  $i,j=1,\dots,n$ 
        set  $d(i,j,k) = \min\{d(i,j,k-1), d(i,k,k-1)+d(k,j,k-1)\}$ 
```

Running time $O(n^3)$

Traveling Salesperson Problem

Given: n cities, and the pairwise distances d_{ij}

Goal: find shortest tour that visits every city at least once

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Observation:

if shortest tour for $T(v, S)$ visits city u right before v , then

$$T(v, S) = T(u, S - v) + d_{uv}$$

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Goal: find shortest tour that visits every city at least once

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Algorithm:

```
for  $v=1, \dots, n$ 
```

```
    set  $T(v, \{v\}) = 0$ 
```

```
for  $k=2, \dots, n$ 
```

```
    for all sets of cities  $S, |S|=k$ 
```

```
        for all  $v$  in  $S$ 
```

```
            set  $T(v, S) = \min_{u \in S-v} T(u, S-v) + d_{uv}$ 
```

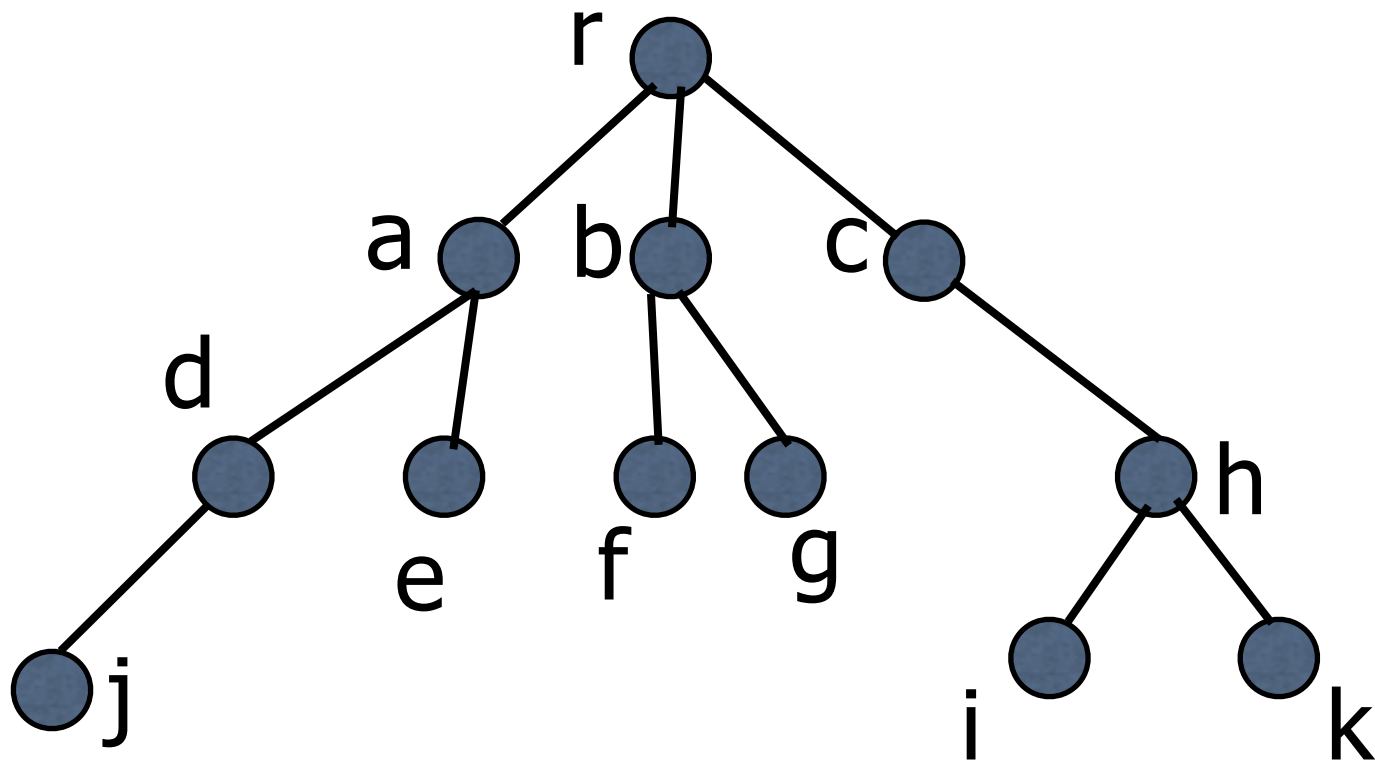
Running time

$O(n^2 2^n)$

Vertex Cover on Acyclic Graphs

Given: A tree

Goal: find smallest vertex cover (vertices that touch all edges)

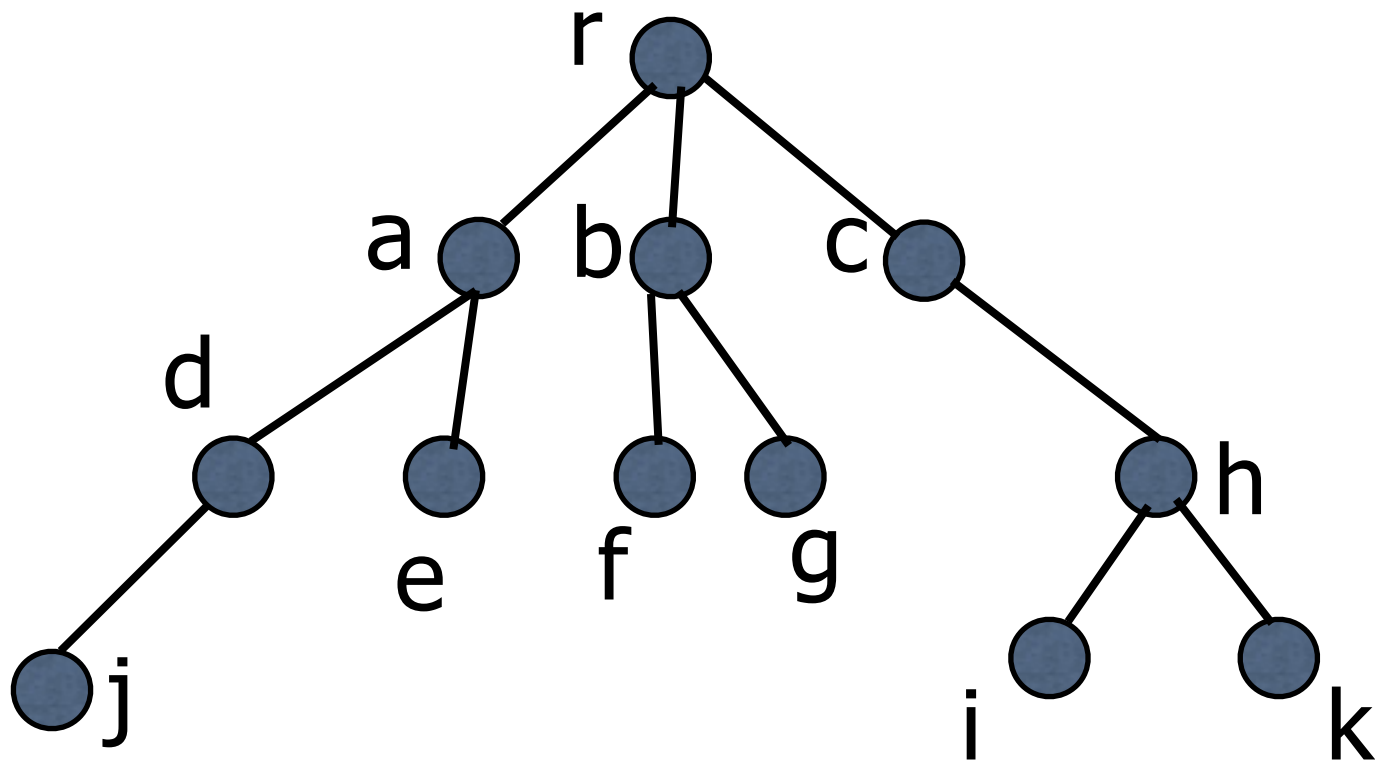


Vertex Cover on Acyclic Graphs

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Subproblems: $V(q)$ - size of smallest vertex cover of subtree rooted at q .



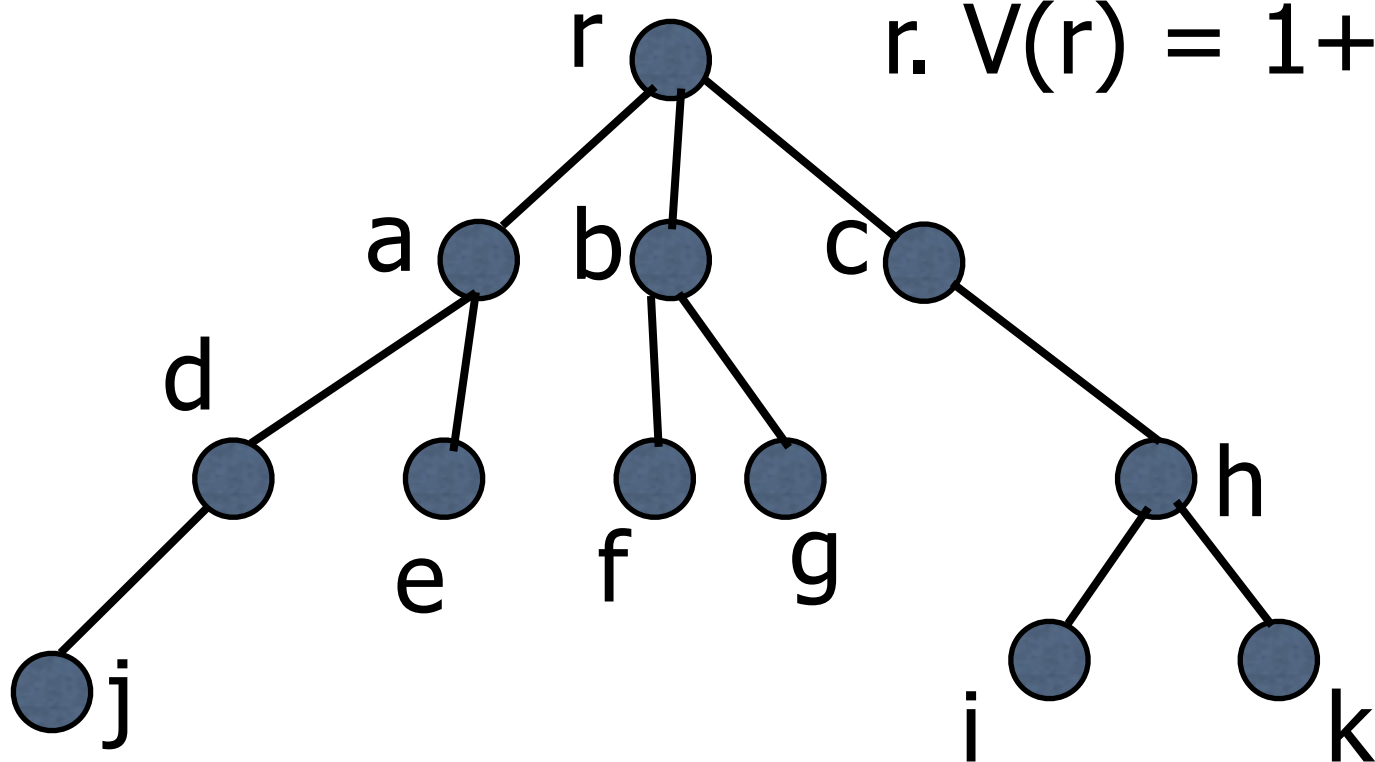
Vertex Cover on Acyclic Graphs

Subproblems: $V(r)$ - size of vertex cover at subtree rooted at r .

Case 1: Cover realizing $V(r)$ does not contain r . Then it must contain $\text{children}(r)$.

$$V(r) = \# \text{children}(r) + \text{sum over grandchildren } g \ V(g)$$

Case 2: Cover realizing $V(r)$ does contain r . $V(r) = 1 + \text{sum over children } c \ V(c)$



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Rough Algorithm:

Running time

$O(n)$

For each vertex r , in decreasing order of depth, set

$$V(r) = \min\{\# \text{children}(r) + \sum_{g, \text{ grandchild of } r} V(g), 1 + \sum_{c, \text{ child of } r} V(c)\}$$

g , grandchild of r

c , child of r

Chain Matrix Multiplication

Given: n matrices $M_1, M_2, \dots, M_n$

Goal: compute product $M_1, M_2, \dots, M_n$ (in what order should we multiply?)

Example: To compute $VWXYZ$ we could multiply $V((WX)(YZ))$ or $(V(W(XY)))Z$ or ...

Basic operations: multiplying $(a \text{ by } b)$ matrix with $(b \text{ by } c)$ matrix gives $(a \text{ by } c)$ matrix in abc time.

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Subproblems: $C(i, j)$ - time to compute $M_i M_{i+1} \dots M_j$

Given: n matrices $M_1, \dots, M_n$, i 'th matrix of size $(m_i \text{ by } m_{i+1})$

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Subproblems: $C(i, j)$ - time to compute $M_i M_{i+1} \dots M_j$

Observation: If the final multiplication in optimal solution is between $(M_i \dots M_k)(M_{k+1} \dots M_j)$, then

$$C(i, j) = C(i, k) + C(k, j) + n_i n_{k+1} n_j .$$

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 $C(i,j) = C(i,k) + C(k+1,j) + m_i m_{k+1} m_j$.

Algorithm:

Running time

$O(n^3)$

for $i=1,2,\dots,n-1$, set $C(i,i)=0$

for $s=1,2,\dots,n-1$, $i=1,\dots,n-1$

set $C(i,i+s) = \min_{i < k < i+s} C(i,k) + C(k+1,i+s) + m_i m_{k+1} m_{i+s}$

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