

Lecture 22

COL 351: Analysis and Design of Algorithms

Lecture 22

Transitivity Relation
 $\text{If } i \rightarrow j \text{ \& } j \rightarrow k \text{ then } i \rightarrow k$
↑ This relation holds for
Reachability property

Matrix Product / All-Pairs Shortest Path / Transitive Closure

(Divide and Conquer strategy)

Transitive Closure

Given: Directed graph $G = (V, E)$.

Find: For each $v \in V$, the vertices reachable from v in graph G .

Naive Way: Run BFS / DFS algorithm from every vertex

- Run time: $O(mn) = O(n^3)$

Better than n^3 algo?

* Floyd Warshall Algo (DP)
 $O(n^3)$

All Pairs Shortest Path

Given: Directed / undirected graph $G = (V, E)$.

Find: The shortest path (or distances) between all vertex pairs.

— Dijkstra's algorithm from every vertex

- Run time: $O(mn + n^2 \log n)$

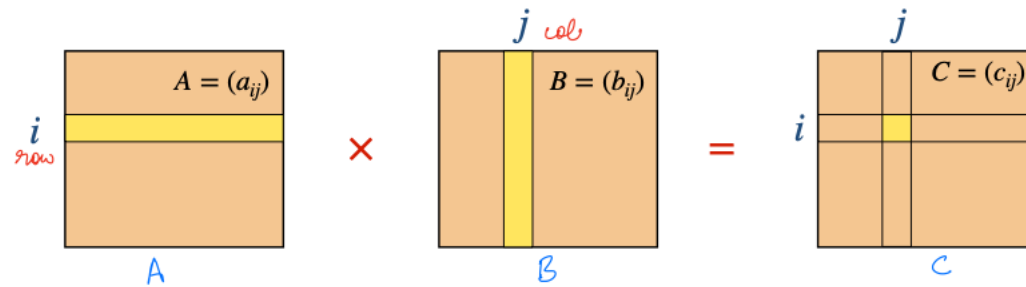
— Floyd-Warshall algorithm (Lec 14)

- Run time: $O(n^3)$

} $O(n^3)$

Better than n^3 algo?

Matrix Multiplication of $n \times n$ Square Matrices



$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$$

Run time = $O(n^3)$

Better than n^3 algo?

Matrix Multiplication

(Divide and Conquer strategy)

Product of 2x2 integer matrices

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \times \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix} = \begin{pmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{pmatrix}$$

8 integer multiplications
4 integer additions

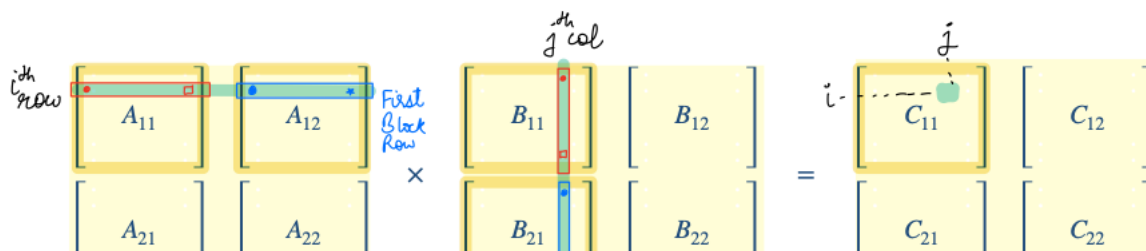
$$c_{11} = a_{11}b_{11} + a_{12}b_{21}$$

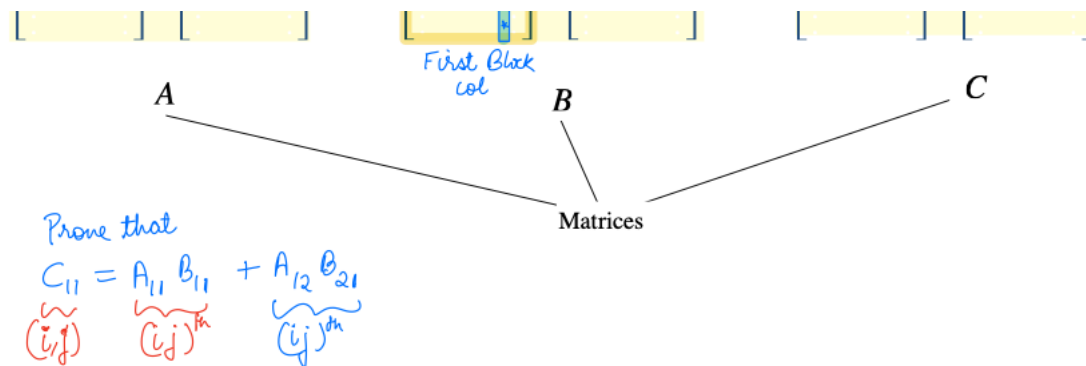
$$c_{12} = \dots a_{11}b_{12} + a_{12}b_{22}$$

$$c_{21} = \dots a_{21}b_{11} + a_{22}b_{21}$$

$$c_{22} = \dots a_{21}b_{12} + a_{22}b_{22}$$

Product of Block-matrices





Product of Block-matrices

$$\begin{array}{c} i^{th} \text{ row} \\ \left[\begin{array}{c|c} A_{11} & A_{12} \\ \hline A_{21} & A_{22} \end{array} \right] \times \begin{array}{c} j^{th} \text{ column} \\ \left[\begin{array}{c|c} B_{11} & B_{12} \\ \hline B_{21} & B_{22} \end{array} \right] = \begin{array}{c} j \\ \left[\begin{array}{c|c} C_{11} & C_{12} \\ \hline C_{21} & C_{22} \end{array} \right] \end{array}$$

$$C_{11} = A_{11}B_{11} + A_{12}B_{21}$$

$$C_{12} = \dots A_{11}B_{12} + A_{12}B_{22}$$

$$C_{21} = \dots A_{21}B_{11} + A_{22}B_{21}$$

$$C_{22} = \dots A_{21}B_{12} + A_{22}B_{22}$$

$$T(n) = 8T(n/2) + O(n^2)$$

$$\text{H.W. } T(n) = ?$$

Strassen's Divide & Conquer Algorithm

$$\begin{array}{c} \left[\begin{array}{c|c} A_{11} & A_{12} \\ \hline A_{21} & A_{22} \end{array} \right] \times \begin{array}{c} \left[\begin{array}{c|c} B_{11} & B_{12} \\ \hline B_{21} & B_{22} \end{array} \right] = \begin{array}{c} \left[\begin{array}{c|c} C_{11} = A_{11}B_{11} + A_{12}B_{21} & C_{12} = A_{11}B_{12} + A_{12}B_{22} \\ \hline C_{21} = A_{21}B_{11} + A_{22}B_{21} & C_{22} = A_{21}B_{12} + A_{22}B_{22} \end{array} \right] \end{array}$$

It suffices
to compute
7 product
matrices

$$\begin{aligned} P_1 &= (A_{11}) \times (B_{12} - B_{22}) \\ P_2 &= (A_{11} + A_{12}) \times (B_{22}) \\ P_3 &= (A_{21} + A_{22}) \times (B_{11}) \\ P_4 &= (A_{22}) \times (B_{21} - B_{11}) \\ P_5 &= (A_{11} + A_{22}) \times (B_{11} + B_{22}) \\ P_6 &= (A_{12} - A_{22}) \times (B_{21} + B_{22}) \\ P_7 &= (A_{11} - A_{21}) \times (B_{11} + B_{12}) \end{aligned}$$

$$\begin{aligned} C_{11} &= (P_4 + P_5 + P_6 - P_2) \\ C_{12} &= (P_1 + P_2) \\ C_{21} &= (P_3 + P_4) \\ C_{22} &= (P_1 + P_5 - P_3 - P_7) \end{aligned}$$

↑
Goal

$$T(n) = 7T(n/2) + O(n^2)$$

Recurrence relation in Strassen's Algorithm for matrix-product is
 $T(n) = 7T(n/2) + d n^2$

$$a^{\log_2(b)} = b^{\log_2(a)}$$

$$T(n) \leq 7T\left(\frac{n}{2}\right) + d n^2$$

$$\leq 7 \left(7T\left(\frac{n}{4}\right) + d \frac{n^2}{4} \right) + d n^2 = 7^2 T\left(\frac{n}{4}\right) + d n^2 \left(1 + \frac{7}{4}\right)$$

$$\leq 7^2 \left(7T\left(\frac{n}{8}\right) + d \frac{n^2}{16} \right) + d n^2 \left(1 + \frac{7}{4}\right)$$

$$\leq 7^3 T\left(\frac{n}{8}\right) + d n^2 \left(1 + \frac{7}{4} + \left(\frac{7}{4}\right)^2\right)$$

⋮

$$\leq 7^i T\left(\frac{n}{2^i}\right) + d n^2 \left(1 + \frac{7}{4} + \left(\frac{7}{4}\right)^2 + \dots + \left(\frac{7}{4}\right)^{i-1}\right)$$

Put $i = \log_2 n$

$$\leq 7^i T\left(\frac{n}{2^i}\right) + \frac{7}{3} d n^2 \left(\frac{7}{4}\right)^i$$

$$\leq 7^{\log_2 n} + \frac{7}{3} d n^2 \left(\frac{7}{4}\right)^{\log_2 n} = O\left(n^{\log_2 7} + n^2 n^{\log_2 7/4}\right) = O\left(n^{\log_2 7}\right)$$

Matrix Multiplication Algorithms

Strassen - $O(n^{2.81})$

⋮

Connersmith and Winograd - $O(n^{2.376})$

Alman and Williams - $O(n^{2.373})$

Notation " ω " is used to denote the "smallest" constant such that two $n \times n$ square matrices can be multiplied in $O(n^\omega)$ time.

$$\omega \geq 2$$

Transitive Closure

Given: Directed graph $G = (V, E)$.

Find: For each $v \in V$, the vertices reachable from v in graph G .

A is an **adjacency-matrix** of G

$$\text{if } A_{ij} = \begin{cases} 1 & \text{if } (i, j) \text{ is edge in } G \\ 0 & \text{otherwise} \end{cases}$$

T is an **transitive-closure-matrix** of G

$$\text{if } T_{ij} = \begin{cases} 1 & \text{there is a path from } (i) \text{ to } (j) \text{ in } G \\ 0 & \text{otherwise} \end{cases}$$

Lemma 1: Let A be adjacency matrix of a graph. Then,

$(A^k)_{ij} > 0$ iff there is a walk of length **exactly** ' k ' from (i) to (j) .

if \exists a walk from i to j of size k

$\text{Hyp}(k)$

$$\text{if } (A^k)_{ij} > 0$$

$$\Rightarrow (A^k)_{ij} > 0$$

H.W.

(Need to use fact that
we are dealing with matrices
of non-negative entries)

$$\Rightarrow \exists \text{ a walk from } i \text{ to } j \text{ of size } k$$

Assume Hyp (k-1) holds.

Now suppose for some i, j , $(A^k)_{ij} > 0$.

$$\Rightarrow \exists x \in [1, n] \text{ s.t. } (A^{k-1})_{ix}, (A)_{xj} > 0$$

By Hyp (k-1), \exists a walk of size (k-1) from
 i to x and also (x, j) is an edge

$$\Rightarrow \exists \text{ a walk of size } k \text{ from } i \text{ to } j.$$

Lemma 1: Let A be adjacency matrix of a graph. Then,

$(A^k)_{ij} > 0$ iff there is a walk of length **exactly** ' k ' from (i) to (j) .

Lemma 2: Let A be adjacency matrix of a graph. Then,

$((I + A)^k)_{ij} > 0$ iff there is a walk of length **at most** ' k ' from (i) to (j) .

Proof Sketch:

$$(I + A)^k = I + \overset{A^0}{\nwarrow} ({}^k C_1)A + ({}^k C_2)A^2 + \dots + A^k$$

$$((I + A)^k)_{ij} > 0 \Leftrightarrow \exists x \in [0, k] \text{ s.t. } (A^x)_{ij} > 0 \Leftrightarrow \exists x \in [0, k] \text{ s.t. there is a walk of size "x" from } i \text{ to } j$$



\exists a path from
 i to j of size $\leq k$

Transitive Closure

Transitive-Closure(A)

$$M = I + A;$$

$$M = M \cdot A;$$


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 $n = \text{size}(A);$   
  
For  $i = 1$  to  $\lceil \log_2 n \rceil$ :  
     $M = M^2$   
    Replace non-zero entries in  $M$  by 1;  
  
Return  $M$ ;
```

Result: The transitive closure of graph with n vertices is computable in $O(n^\omega \log n)$ time.