

Today's class: All Pair reachability in $O(n^2)$ time in fully dynamic setting in DAGs. with $O(n)$ query time.

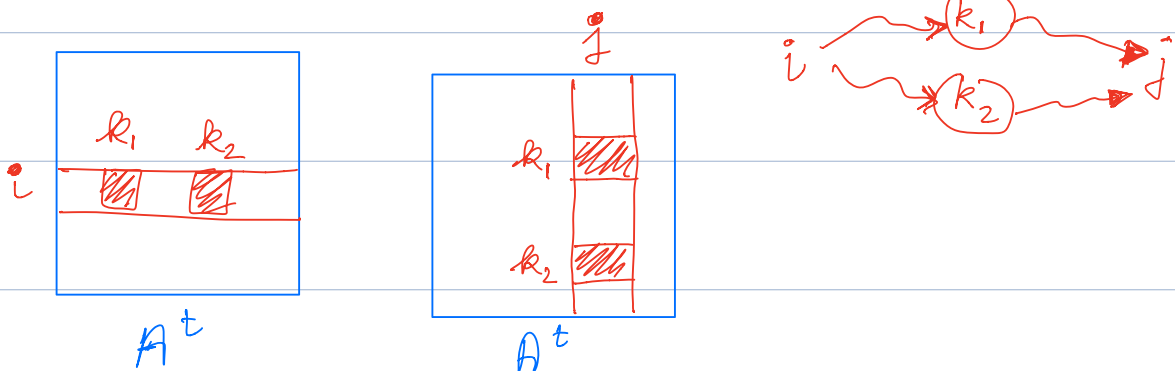
Static All-Pairs Reachability

$$A^t(x, y) = \begin{cases} \neq 0 & \text{if } \exists \text{ } x \text{ to } y \text{ path in } G \\ & \text{of length at most } t. \\ 0 & \text{o/w.} \end{cases}$$

What is A^1 ?? (Adj + I)

How to compute A^{2t} from A^t ?

Solⁿ: Take A^t , make nonzero entries of it 1.
Output $(A^t)^* (A^t)$.



Claim: Time to find A^n from A^1 is $O(n^{\log n})$, as we do $\log n$ multiplications to reach A^n from A^1 .

Theorem : Time to multiply $n \times B$ & $B \times n$ matrix is $O\left(\frac{n^2}{B^{2-\omega}}\right)$ $\omega \leq 2.38$
matrix mult coeff

Proof :

M_1

$B \times B$

M_2

$B \times B$

M_3

$B \times B$

$\left[\begin{array}{ccc} \tilde{M}_1 & \tilde{M}_2 & \tilde{M}_3 \end{array} \right]$

$\tilde{M}_1$

$B \times B$

$\tilde{M}_2$

$B \times B$

$\tilde{M}_3$

$B \times B$

$B \times n$

$n \times B$

(assume $\frac{n}{B} = \text{integer}$)

$=$

$M_1, \tilde{M}_1$

$M_2, \tilde{M}_1$

$M_3, \tilde{M}_1$

$\left[\begin{array}{ccc} M_1, \tilde{M}_1 & M_1, \tilde{M}_2 & M_1, \tilde{M}_3 \\ M_2, \tilde{M}_1 & M_2, \tilde{M}_2 & M_2, \tilde{M}_3 \\ M_3, \tilde{M}_1 & M_3, \tilde{M}_2 & M_3, \tilde{M}_3 \end{array} \right]$

$M_1, \tilde{M}_2$

$M_2, \tilde{M}_2$

$M_3, \tilde{M}_2$

$\left[\begin{array}{ccc} M_1, \tilde{M}_3 & M_2, \tilde{M}_3 & M_3, \tilde{M}_3 \end{array} \right]$

$M_1, \tilde{M}_3$

$M_2, \tilde{M}_3$

$M_3, \tilde{M}_3$

$n \times n$

So time = $\left(\frac{n}{B}\right)^2 \cdot O(B^\omega) = O\left(\frac{n^2}{B^{2-\omega}}\right)$

← x → x →

Theorem: Let $P_1, P_2, \dots, P_B$ be B columns of size $n \times 1$ and $Q_1, Q_2, \dots, Q_B$ are B rows of size $1 \times n$, then

$$\sum_{j=1}^B (P_j \cdot Q_j) = [P_1 \dots P_B] \cdot [Q_1 \dots Q_B]$$

$$\sum_{j=1}^B \left(\begin{array}{c} P_j \end{array} \cdot \begin{array}{c} Q_j \end{array} \right) = \left(\begin{array}{c} P_1 \dots P_B \end{array} \right) \cdot \left(\begin{array}{c} Q_1 \\ \vdots \\ Q_B \end{array} \right)$$

$B \times n$

$n \times B$

Proof Hint:

It suffices to show that

$(i, k)^{th}$ entry of $\sum_{j=1}^B (P_j \cdot Q_j)$ is same as

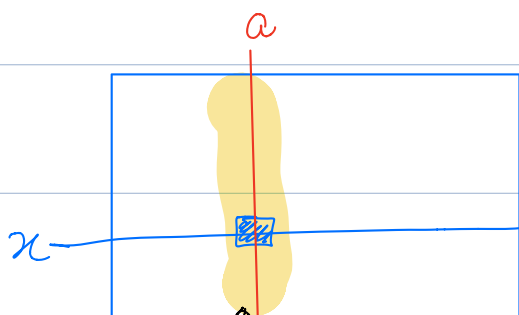
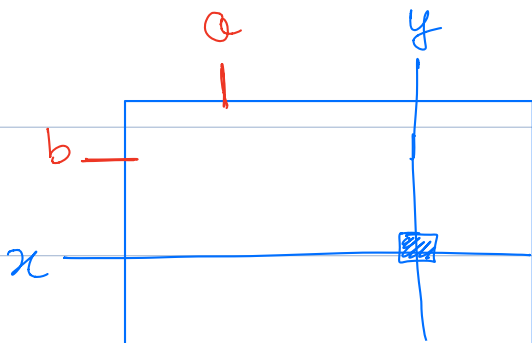
$(i, k)^{th}$ entry of $[P_1 \dots P_B] \cdot [Q_1 \dots Q_B]$.

————— x ——— x —————

Ques: Can we get better than n^2 update time at cost of higher query time?
 (We will update $\tilde{C}$ in lazy manner).

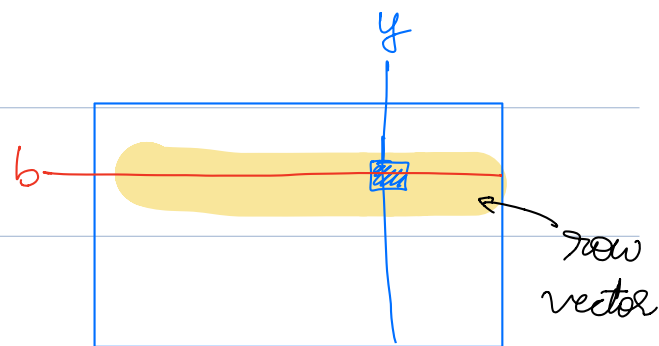
Algorithm 2 :

Recall $C_t(x, y) = C_{t-1}(x, y) \pm C_{t-1}(x, a) \cdot C_{t-1}(b, y)$
 ↑
 Addition/deletion of (a, b)



column vector $(C_{t-1}^{col}(a))$

*



row vector $(C_{t-1}^{row}(b))$

OBSERVATION:

Take product of $C_{t-1}^{\text{col}}(a)$ & $C_{t-1}^{\text{row}}(b)$.

The $(x, y)^{\text{th}}$ entry here is the term to be $C_{t-1}(x, a) \cdot C_{t-1}(b, y)$ added or subtracted.

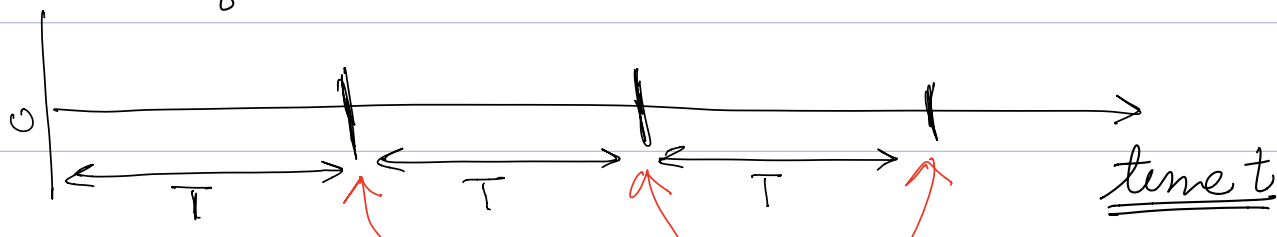
Fact: $C_t = C_{t-1} \overset{\pm}{\pm} \underbrace{C_{t-1}^{\text{col}}(a)}_{\text{column vector}} \cdot \underbrace{C_{t-1}^{\text{row}}(b)}_{\text{row vector}}$

addⁿ / delⁿ of edge (a, b)

So, if (a_1, b_1) (a_2, b_2) ... (a_T, b_T) are updates then,

$$C_T = C_0 + \underbrace{\begin{bmatrix} \pm C_0^{\text{col}}(a_1) & \pm C_1^{\text{col}}(a_2) & \dots & \pm C_{T-1}^{\text{col}}(a_T) \end{bmatrix}}_{n \times T} \cdot \underbrace{\begin{bmatrix} C_0^{\text{row}}(b_1) \\ C_1^{\text{row}}(b_2) \\ \vdots \\ C_{T-1}^{\text{row}}(b_T) \end{bmatrix}}_{T \times n}$$

Batch of T updates:



Update the matrix " $\tilde{C}$ "
only after batch of T edge updates.

How to answer queries ??

Suppose we have " C_0 ", during a batch of T updates we explicitly maintain:

$$C_0^{\text{col}}(a_1) \dots C_{T-1}^{\text{col}}(a_T) \text{ \& } C_0^{\text{row}}(b_1) \dots C_{T-1}^{\text{row}}(b_T)$$

After $K \leq T$ edge updates if query (x, y) comes then

$$C_K(x, y) = C_0(x, y) + \sum_{i=1}^K \pm \underbrace{C_{i-1}^{\text{col}}(a_i)[x]}_{(x, a_i)} \cdot \underbrace{C_{i-1}^{\text{row}}(b_i)[y]}_{(b_i, y)}$$

So, query time is $O(K)$
 $= O(T)$

• Now the question is how to maintain $C_0^{\text{col}}(a_1) \dots C_{T-1}^{\text{col}}(a_T)$ & $C_0^{\text{row}}(b_1) \dots C_{T-1}^{\text{row}}(b_T)$

• After K updates, we would have computed $C_0^{\text{col}}(a_1) \dots C_{K-1}^{\text{col}}(a_K)$ & $C_0^{\text{row}}(b_1) \dots C_{K-1}^{\text{row}}(b_K)$

• Now at $(K+1)^{\text{th}}$ edge update, i.e. (a_{K+1}, b_{K+1}) added/removed we need to find $C_K^{\text{col}}(a_{K+1})$ & $C_K^{\text{row}}(b_{K+1})$.

i.e. $\forall v, C_K(v, a_{K+1}) \neq C_K(b_{K+1}, v)$.

As there are $2n$ entries to update we can just spend $O(T)$ time for all $2n$ values.

Total time for this task after each edge update $= O(Tn)$.

Now amortized update time is

$$O(nT) + \underbrace{O\left(\frac{n^2}{T^{2-\omega}}\right)}_T$$

$$= O\left(nT + \frac{n^2}{T^{3-\omega}}\right)$$

$$nT = \frac{n^2}{T^{3-\omega}} \Rightarrow T = n^{\frac{1}{4-\omega}}$$

$$= O\left(n^{1 + \frac{1}{4-\omega}}\right) = O(n^{1.616})$$

$$\text{Query time} = O(T) = O(n^{0.616}).$$