



max planck institut
informatik

Complexity Theory of Polynomial-Time Problems

Lecture 8: (Boolean) Matrix Multiplication

Karl Bringmann

Recall: Boolean Matrix Multiplication

given $n \times n$ matrices A, B with entries in $\{0,1\}$

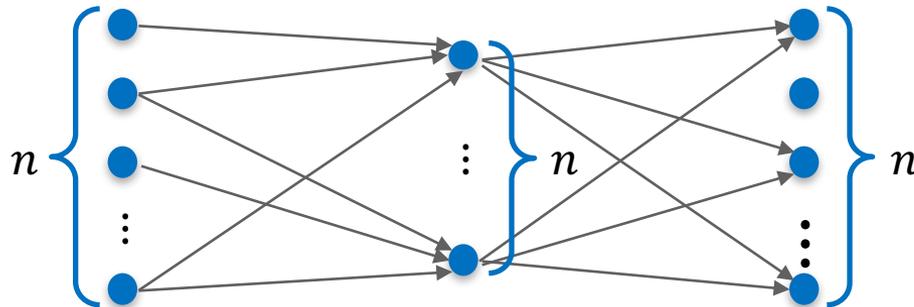
compute matrix C with $C_{i,j} = \bigvee_{k=1}^n A_{i,k} \wedge B_{k,j}$

what we already know about BMM:

BMM is in time $O(n^3 / \log n)$ (four Russians)

BMM is equivalent to computing the **Transitive Closure** of a given graph

BMM can be reduced to APSP $\rightarrow O(n^3 / 2^{\sqrt{\log n}})$



Exponent of Matrix Multiplication

define ω as the *infimum* over all c such that MM has an $O(n^c)$ algorithm

note: MM is in time $O(n^{\omega+\varepsilon})$ for any $\varepsilon > 0$

we will be sloppy and write: MM is in time $O(n^\omega)$

note: MM is **not** in time $O(n^{\omega-\varepsilon})$ for any $\varepsilon > 0$

$$\omega \geq 2$$

Thm:

$$\omega < 3$$

$\omega \leq \dots$

this is very fast – in theory

all these algorithms have
impractically large constant factors

(maybe except Strassen'69)

Strassen'69	2.81
Pan'78	2.79
Bini et al.'79	2.78
Schönhage'80	2.52
Romani'80	2.52
Coppersmith, Winograd'81	2.50
Strassen'86	2.48
Coppersmith, Winograd'90	2.376
Stothers'10	2.374
Vassilevska-Williams'11	2.37288
Le Gall'14	2.37287



Boolean Matrix Multiplication

Thm: BMM is in time $O(n^\omega)$

given $n \times n$ matrices A, B with entries in $\{0,1\}$

compute standard matrix product C' with $C'_{i,j} = \sum_{k=1}^n A_{i,k} \cdot B_{k,j}$

define matrix C with $C_{i,j} = [C'_{i,j} > 0]$

then C is the Boolean matrix product of A and B

Hypothesis: BMM is not in time $O(n^{\omega-\varepsilon})$



Combinatorial Algorithms

fast matrix multiplication uses algebraic techniques which are impractical

“combinatorial algorithms”: do not use algebraic techniques

not well defined!

Arlazarov,Dinic,Kronrod,
Faradzhev'70 (four russians)

$$O(n^3 / \log^2 n)$$

Bansal,Williams'09

$$O(n^3 (\log \log n)^2 / \log^{9/4} n)$$

Chan'15

$$O(n^3 (\log \log n)^3 / \log^3 n)$$

Yu'15

$$O(n^3 \text{ poly } \log \log n / \log^4 n)$$

Hypothesis: BMM has no “combinatorial” algorithm in time $O(n^{3-\varepsilon})$



In this lecture you learn that...

...(B)MM is useful for **designing theoretically fast algorithms**

- Exercise: k-Clique in $O(n^{\omega k/3})$
- Exercise: MaxCut in $O(2^{\omega n/3} \text{poly}(n))$
- Node-Weighted Negative Triangle in $O(n^\omega)$

...BMM is an **obstacle for practically fast / theoretically very fast algorithms**

no combinatorial $O(n^{3-\epsilon})$

not faster than $O(n^{2.373})$

- Transitive Closure has no $O(n^{\omega-\epsilon})$ / combinatorial $O(n^{3-\epsilon})$ algorithm
- Exercise: pattern matching with 2 patterns
- Sliding Window Hamming Distance
- context-free grammar problems



Outline

1) Relations to Subcubic Equivalences

2) Strassen's Algorithm

3) Sliding Window Hamming Distance

4) Node-Weighted Negative Triangle

5) Context-Free Grammars



Corollaries from Subcubic Equivalences

BMM



All-Pairs-Triangle

given an unweighted graph G
vertices $V = I \cup J \cup K$

$\forall i, j$: are they in a triangle with some k ?



Triangle

given an unweighted graph G
does it contain a triangle?

APSP



Min-Plus Product



All-Pairs-Negative-Triangle

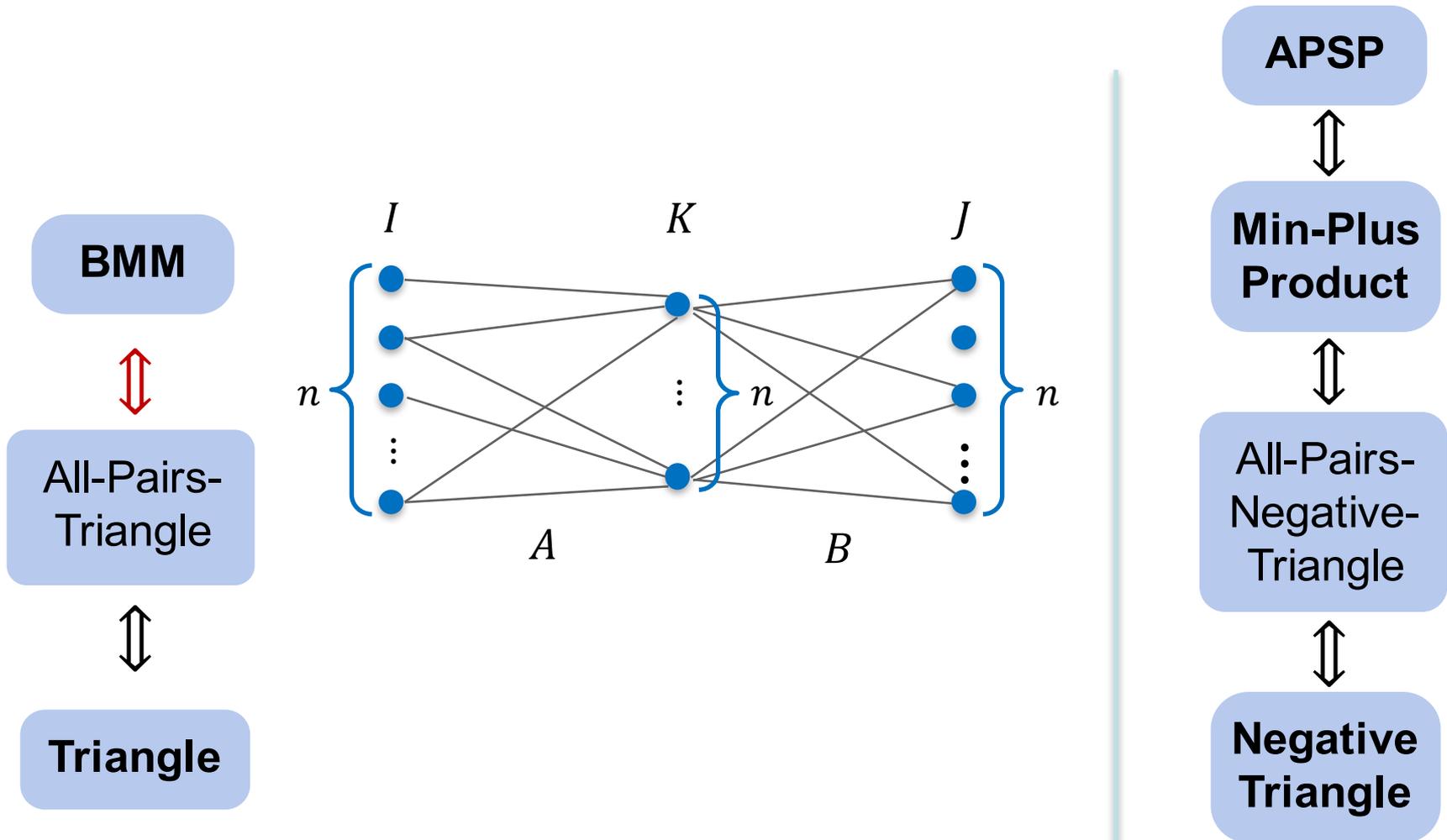


Negative Triangle

[Vassilevska-Williams, Williams'10]



Corollaries from Subcubic Equivalences



Corollaries from Subcubic Equivalences

Given an unweighted undirected graph G

Adjacency matrix A , entries in $\{0,1\}$

BMM



All-Pairs-Triangle



Triangle

1. Compute Boolean Product $C := A * A$:

$$C_{i,j} = \bigvee_k A_{i,k} \wedge A_{k,j}$$

2. Compute $\bigvee_{i,j} A_{i,j} \wedge C_{i,j}$

this equals $\bigvee_{i,j,k} A_{i,j} \wedge A_{i,k} \wedge A_{k,j}$

thus we solved triangle detection

APSP



Min-Plus Product



All-Pairs-Negative-Triangle



Negative Triangle



Corollaries from Subcubic Equivalences



All-Pairs-Triangle to Triangle

All-Pairs-Triangle



Triangle

Triangle Given graph G

Decide whether there are vertices i, j, k such that i, j, k form a triangle

All-Pairs-Triangle Given graph G with vertex set $V = I \cup J \cup K$

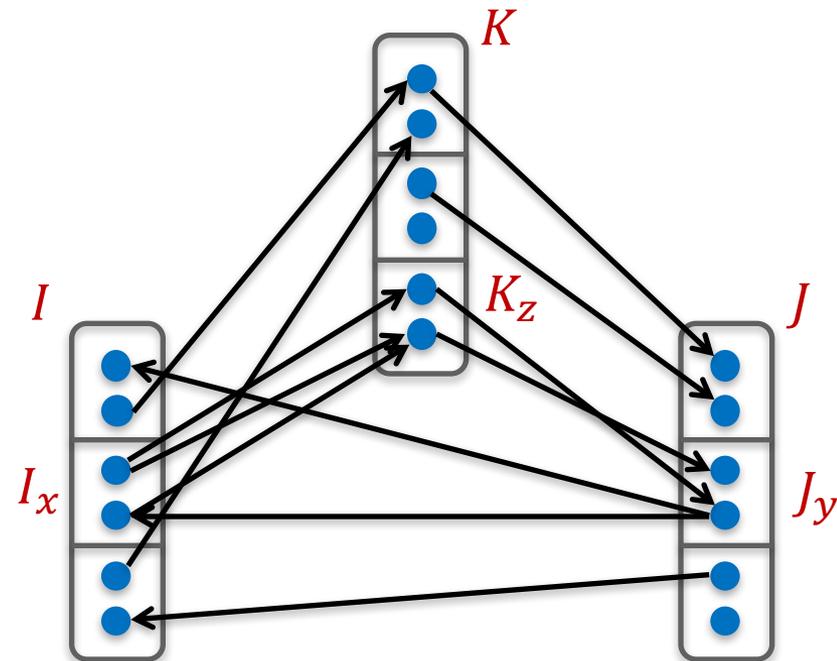
Decide for every $i \in I, j \in J$ whether there is a vertex $k \in K$ such that i, j, k form a triangle

Split I, J, K into n/s parts of size s :

$I_1, \dots, I_{n/s}, J_1, \dots, J_{n/s}, K_1, \dots, K_{n/s}$

For each of the $(n/s)^3$ triples (I_x, J_y, K_z) :

consider graph $G[I_x \cup J_y \cup K_z]$



All-Pairs-Triangle to Triangle

Initialize C as $n \times n$ all-zeroes matrix

For each of the $(n/s)^3$ triples of parts (I_x, J_y, K_z) :

While $G[I_x \cup J_y \cup K_z]$ contains a triangle:

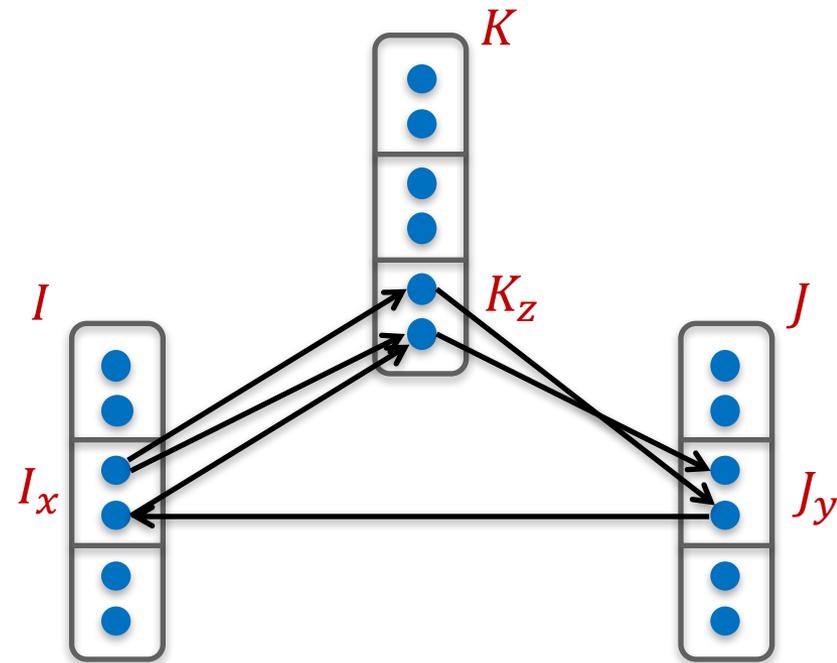
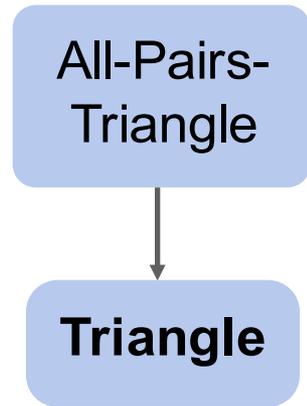
Find a triangle (i, j, k) in $G[I_x \cup J_y \cup K_z]$

Set $C[i, j] := 1$

Delete edge (i, j)

(i, j) is in no more triangles

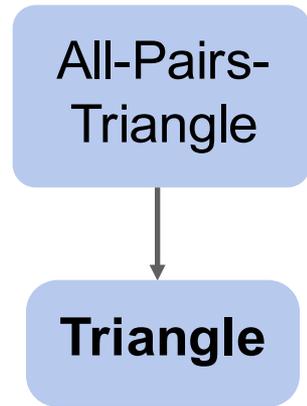
- ✓ guaranteed termination:
can delete $\leq n^2$ edges
- ✓ correctness:
if (i, j) is in a triangle,
we will find one



All-Pairs-Triangle to Triangle

Find a triangle (i, j, k) in $G[I_x \cup J_y \cup K_z]$

How to **find** a triangle
if we can only **decide** whether one exists?



Partition I_x into $I_x^{(1)}, I_x^{(2)}$, J_y into $J_y^{(1)}, J_y^{(2)}$, K_z into $K_z^{(1)}, K_z^{(2)}$

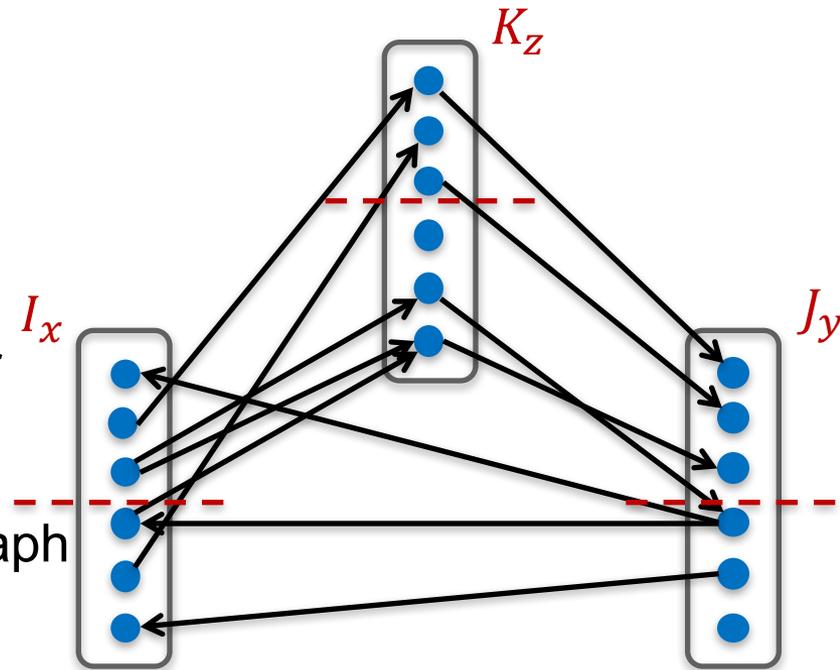
Since $G[I_x \cup J_y \cup K_z]$ contains a triangle,
at least one of the 2^3 subgraphs

$$G[I_x^{(a)} \cup J_y^{(b)} \cup K_z^{(c)}]$$

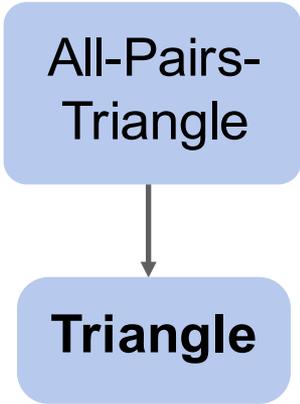
contains a triangle

Decide for each such subgraph whether
it contains a triangle

Recursively find a triangle in one subgraph



All-Pairs-Triangle to Triangle



Find a triangle (i, j, k) in $G[I_x \cup J_y \cup K_z]$

How to **find** a triangle
if we can only **decide** whether one exists?

Partition I_x into $I_x^{(1)}, I_x^{(2)}$, J_y into $J_y^{(1)}, J_y^{(2)}$, K_z into $K_z^{(1)}, K_z^{(2)}$

Since $G[I_x \cup J_y \cup K_z]$ contains a triangle,
at least one of the 2^3 subgraphs

$$G[I_x^{(a)} \cup J_y^{(b)} \cup K_z^{(c)}]$$

contains a triangle

Decide for each such subgraph whether
it contains a triangle

Recursively find a triangle in one subgraph

Running Time:

$$T_{\text{FindTriangle}}(n) \leq$$

$$2^3 \cdot T_{\text{DecideTriangle}}(n)$$

$$+ T_{\text{FindTriangle}}(n/2)$$

$$= O(T_{\text{DecideTriangle}}(n))$$



All-Pairs-Triangle to Triangle

Initialize C as $n \times n$ all-zeroes matrix

For each of the $(n/s)^3$ triples of parts (I_x, J_y, K_z) :

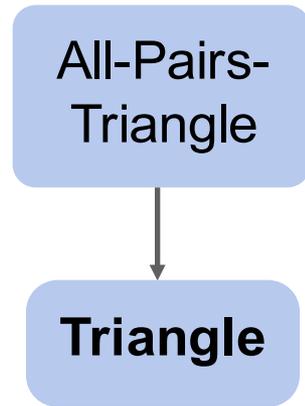
While $G[I_x \cup J_y \cup K_z]$ contains a triangle:

Find a triangle (i, j, k) in $G[I_x \cup J_y \cup K_z]$

Set $C[i, j] := 1$

Delete edge (i, j)

} (*)



Running Time:

$$(*) = O(T_{\text{FindTriangle}}(s)) = O(T_{\text{DecideTriangle}}(s))$$

$$\text{Total time: } ((\# \text{triples}) + (\# \text{triangles found})) \cdot (*)$$

$$\leq ((n/s)^3 + n^2) \cdot T_{\text{DecideTriangle}}(s)$$

$$\text{Set } s = n^{1/3} \text{ and assume } T_{\text{DecideTriangle}}(n) = O(n^{3-\varepsilon})$$

$$\text{Total time: } O(n^2 \cdot n^{1-\varepsilon/3}) = O(n^{3-\varepsilon/3})$$



Corollaries from Subcubic Equivalences

BMM



All-Pairs-Triangle



Triangle

If BMM has (combinatorial) $O(n^{3-\varepsilon})$ algorithm
then Triangle has (combinatorial) $O(n^{3-\varepsilon})$ algorithm

If Triangle has (combinatorial) $O(n^{3-\varepsilon})$ algorithm
then BMM has (combinatorial) $O(n^{3-\varepsilon/3})$ algorithm

→ **subcubic equivalent**,
but this mainly makes sense
for **combinatorial** algorithms

APSP



Min-Plus Product



All-Pairs-Negative-Triangle



Negative Triangle



Outline

1) Relations to Subcubic Equivalences

2) Strassen's Algorithm

3) Sliding Window Hamming Distance

4) Node-Weighted Negative Triangle

5) Context-Free Grammars



Strassen's Algorithm

shows $\omega \leq 2.81$

$$\begin{array}{c|c} & A \\ \hline & B \\ \hline & C \end{array}$$

·	=

The diagram illustrates the matrix multiplication $A \cdot B = C$. Matrix A is a 2x2 grid with elements $A_{1,1}$, $A_{1,2}$, $A_{2,1}$, and $A_{2,2}$. Matrix B is a 2x2 grid with elements $B_{1,1}$, $B_{1,2}$, $B_{2,1}$, and $B_{2,2}$. Matrix C is a 2x2 grid with elements $C_{1,1}$, $C_{1,2}$, $C_{2,1}$, and $C_{2,2}$. The matrices are arranged in a row, separated by a dot and an equals sign.

$$C_{1,1} = A_{1,1} \cdot B_{1,1} + A_{1,2} \cdot B_{2,1}$$

$$C_{1,2} = A_{1,1} \cdot B_{1,2} + A_{1,2} \cdot B_{2,2}$$

$$C_{2,1} = A_{2,1} \cdot B_{1,1} + A_{2,2} \cdot B_{2,1}$$

$$C_{2,2} = A_{2,1} \cdot B_{1,2} + A_{2,2} \cdot B_{2,2}$$

$$T(n) \leq 8T(n/2) + O(n^2)$$

$$T(n) \leq O(n^3)$$



Strassen's Algorithm

shows $\omega \leq 2.81$

$$\begin{array}{c|c} & A \\ \hline & B \\ \hline & C \end{array}$$

$A_{1,1}$	$A_{1,2}$
$A_{2,1}$	$A_{2,2}$

 ·

$B_{1,1}$	$B_{1,2}$
$B_{2,1}$	$B_{2,2}$

 =

$C_{1,1}$	$C_{1,2}$
$C_{2,1}$	$C_{2,2}$

$$M_1 = (A_{1,1} + A_{2,2}) \cdot (B_{1,1} + B_{2,2})$$

$$M_2 = (A_{2,1} + A_{2,2}) \cdot B_{1,1}$$

$$M_3 = A_{1,1} \cdot (B_{1,2} - B_{2,2})$$

$$M_4 = A_{2,2} \cdot (B_{2,1} - B_{1,1})$$

$$M_5 = (A_{1,1} + A_{1,2}) \cdot B_{2,2}$$

$$M_6 = (A_{2,1} - A_{1,1}) \cdot (B_{1,1} + B_{1,2})$$

$$M_7 = (A_{1,2} - A_{2,2}) \cdot (B_{2,1} + B_{2,2})$$

$$C_{1,1} = M_1 + M_4 - M_5 + M_7$$

$$C_{1,2} = M_3 + M_5$$

$$C_{2,1} = M_2 + M_4$$

$$C_{2,2} = M_1 - M_2 + M_3 + M_6$$

$$T(n) \leq 7 T(n/2) + O(n^2)$$

$$T(n) \leq O(n^{\log_2 7}) = O(n^{2.8074})$$



Faster Matrix Multiplication

tensor = 3-dimensional matrix

matrix multiplication tensor:

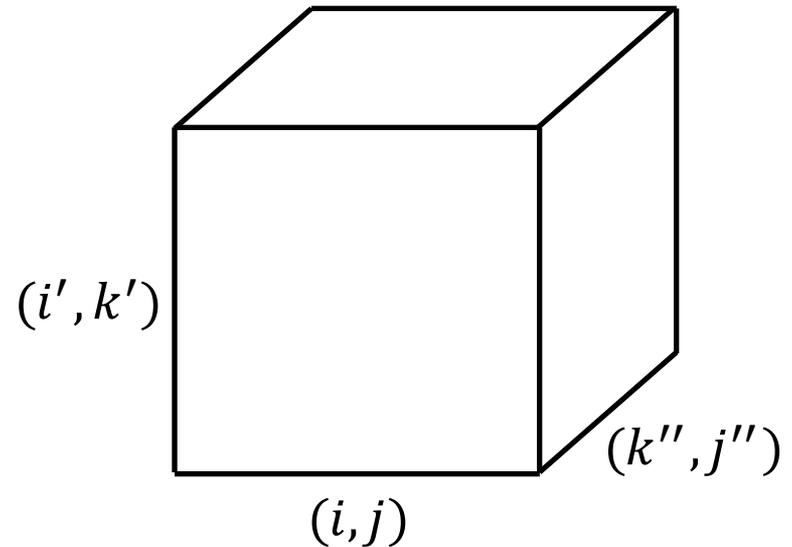
n^2 rows/columns/...

entries in $\{0,1\}$

entry $T_{(i,j),(i',k'),(k'',j'')} = 1$

iff $i = i'$ and $j = j''$ and $k' = k''$

i.e. $A_{i',k'} \cdot B_{k'',j''}$ appears in $C_{i,j}$



matrix of rank 1: outer product of two vectors

matrix of rank r : sum of r rank-1-matrices

tensor of rank 1: outer product of three vectors

tensor of rank r : sum of r rank-1-tensors

matrix rank is in P

tensor rank is not
known to be in P



Faster Matrix Multiplication

tensor = 3-dimensional matrix

matrix multiplication tensor:

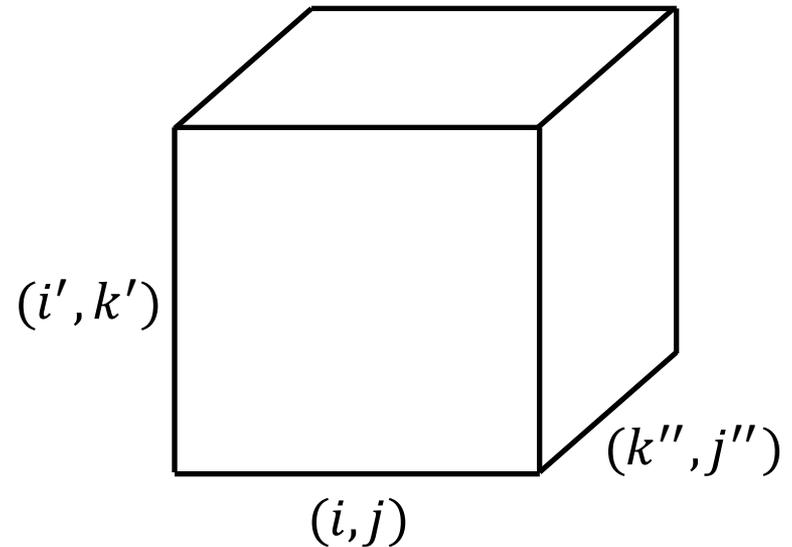
n^2 rows/columns/...

entries in $\{0,1\}$

entry $T_{(i,j),(i',k'),(k'',j'')} = 1$

iff $i = i'$ and $j = j''$ and $k' = k''$

i.e. $A_{i',k'} \cdot B_{k'',j''}$ appears in $C_{i,j}$



Strassen: rank of MM-tensor for $n = 2$ is at most 7

any bound on rank of MM-tensor can be transformed into a MM-algorithm

thus search for faster MM-algorithms is a mathematical question

this is **complete**: one can find ω by analyzing tensor rank!



Outline

- 1) Relations to Subcubic Equivalences
- 2) Strassen's Algorithm
- 3) Sliding Window Hamming Distance**
- 4) Node-Weighted Negative Triangle
- 5) Context-Free Grammars



Sliding Window Hamming Distance

given two strings: text T of length n and pattern P of length $m < n$

compute for each i the Hamming distance of P and $T[i..i + m - 1]$

best known algorithm:

$$O(n \sqrt{m} \text{polylog } n) \\ \leq O(n^{1.5001})$$

	a	b	c	b	b	c	a	a	
	b	b	c	a					2
		b	b	c	a				3
			b	b	c	a			3
				b	b	c	a		0
					b	b	c	a	2

[Indyk, Porat, Clifford'09]

Thm: Sliding Window Hamming Distance has no $O(n^{\omega/2-\epsilon})$ algorithm or combinatorial $O(n^{1.5-\epsilon})$ algorithm unless the BMM-Hypothesis fails

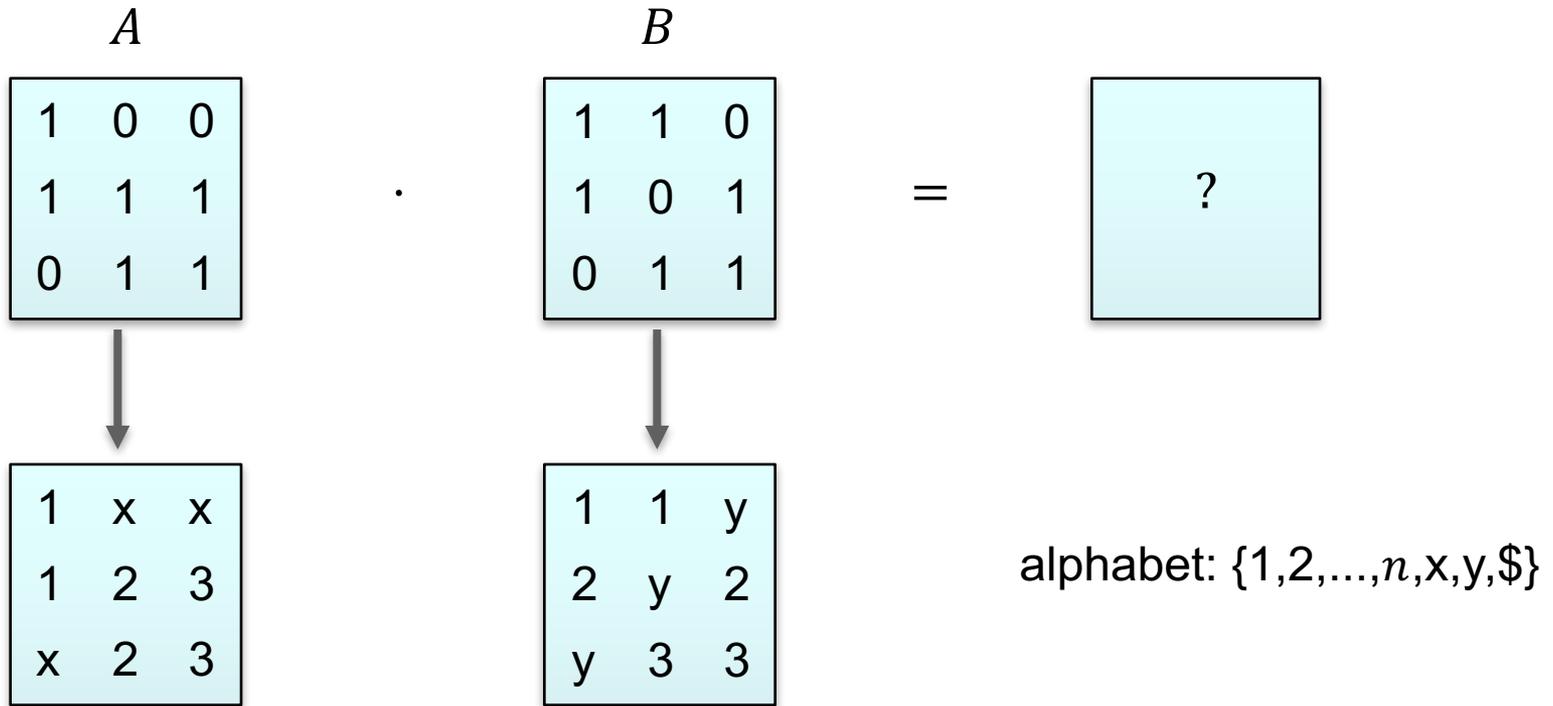
$$\approx O(n^{1.18})$$

Open Problem: get rid of „combinatorial“ or design improved algorithm using MM



Sliding Window Hamming Distance

given two strings: text T of length n and pattern P of length $m < n$
 compute for each i the Hamming distance of P and $T[i..i + m - 1]$



pattern = concat rows:

1 x x 1 2 3 x 2 3

text = concat columns + padding:

\$ \$ \$ \$ \$ \$ 1 2 y \$ 1 y 3 \$ y 2 3 \$ \$ \$ \$ \$ \$

1 x x 1 2 3 x 2 3



Sliding Window Hamming Distance

\$	\$	\$	\$	\$	\$	1	2	y	\$	1	y	3	\$	y	2	3	\$	\$	\$	\$	\$	\$
						1	x	x	1	2	3	x	2	3								
										1	x	x	1	2	3	x	2	3				
														1	x	x	1	2	3	x	2	3
			1	x	x	1	2	3	x	2	3											
							1	x	x	1	2	3	x	2	3							
1	x	x	1	2	3	x	2	3														
				1	x	x	1	2	3	x	2	3										
							1	x	x	1	2	3	x	2	3							

put a 1 if there is at least one match



1	0	0
1	1	1
0	1	1

.

1	1	0
1	0	1
0	1	1

=

1	1	0
1	1	1
1	1	1



Sliding Window Hamming Distance

given Boolean $n \times n$ -matrices A, B

we construct text+pattern of length $O(n^2)$ (in time $O(n^2)$)

thus, an $O(n^{\omega/2-\varepsilon})$ algorithm for Sliding Window Hamming Distance

would yield an $O(n^{\omega-2\varepsilon})$ algorithm for BMM, contradicting BMM-Hypothesis

and an $O(n^{1.5-\varepsilon})$ combinatorial algorithm for Sliding Window Hamming Dist.

would yield an $O(n^{3-2\varepsilon})$ combinatorial algorithm for BMM

Thm: Sliding Window Hamming Distance has no $O(n^{\omega/2-\varepsilon})$ algorithm or combinatorial $O(n^{1.5-\varepsilon})$ algorithm unless the BMM-Hypothesis fails



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- 1) Relations to Subcubic Equivalences
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- 4) Node-Weighted Negative Triangle**
- 5) Context-Free Grammars



Node-Weighted Negative Triangle

(Edge-Weighted) Negative Triangle

$O(n^3)$

given a directed graph with weights $w_{i,j}$ on **edges**,
is there a triangle i, j, k : $w_{j,i} + w_{i,k} + w_{k,j} < 0$?

Node-Weighted Negative Triangle

given a directed graph with weights w_i on **nodes**,
is there a triangle i, j, k : $w_i + w_j + w_k < 0$?

$w_i := -1$

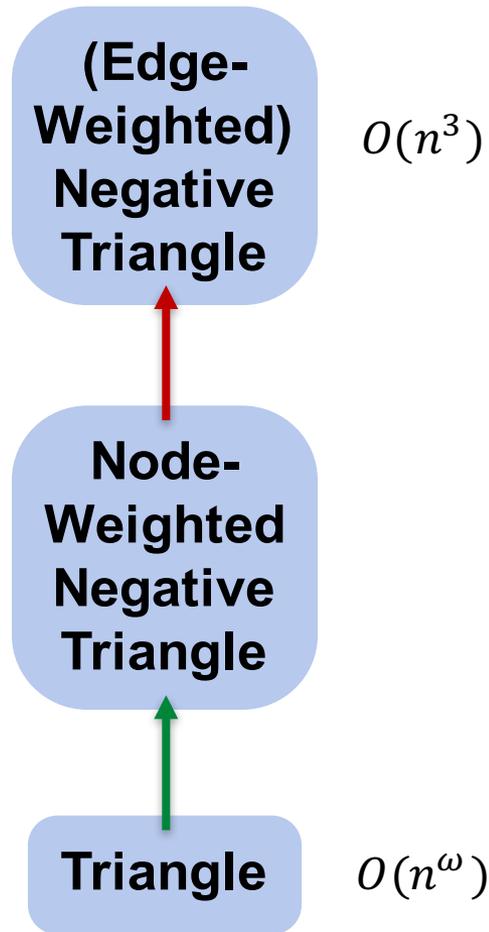
Triangle

$O(n^\omega)$

given an **unweighted** undirected graph,
is there a triangle?



Node-Weighted Negative Triangle



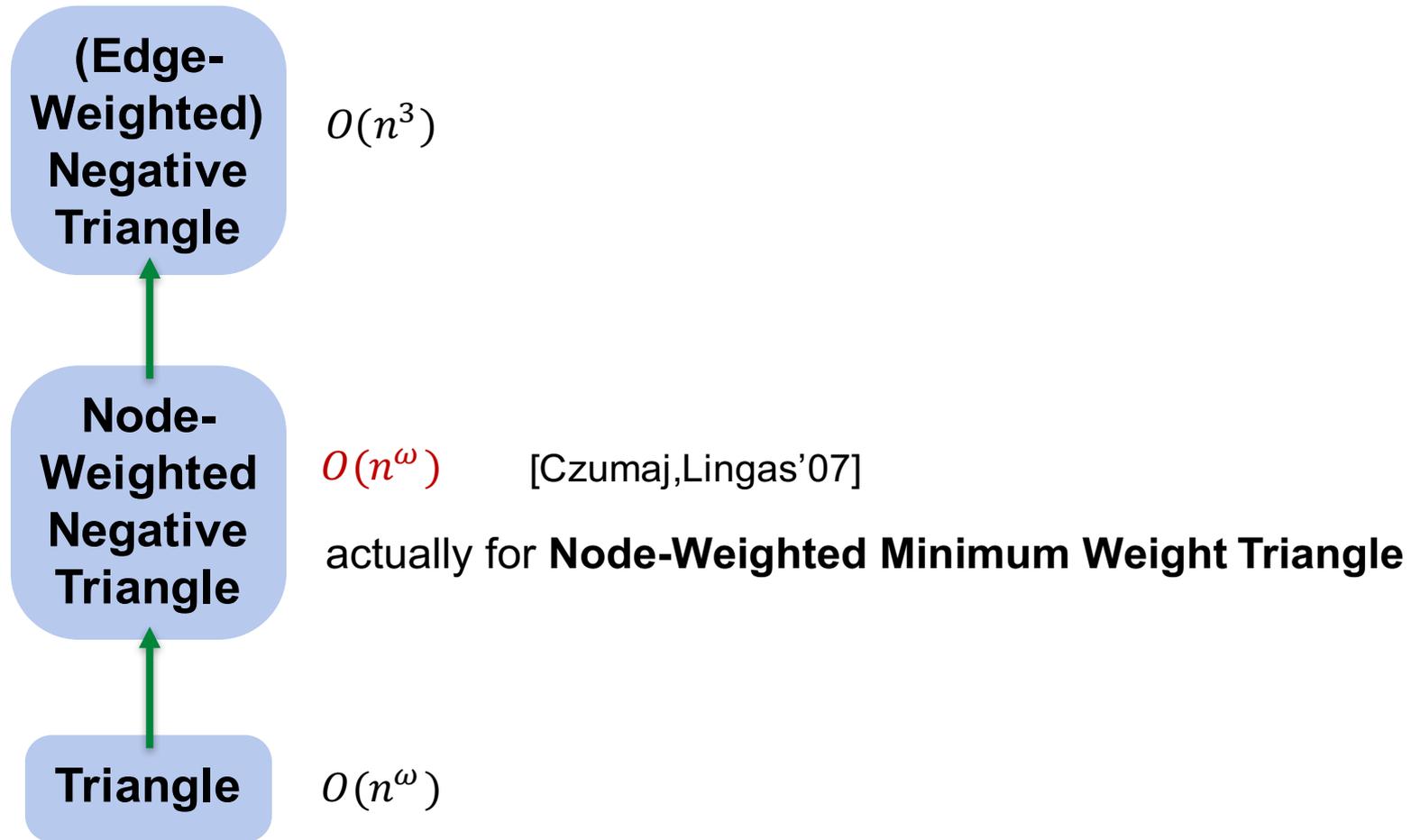
find appropriate **edge** weights that simulate the given **node** weights:

$$\text{set } w_{i,j} := (w_i + w_j)/2$$

then for a triangle i, j, k :

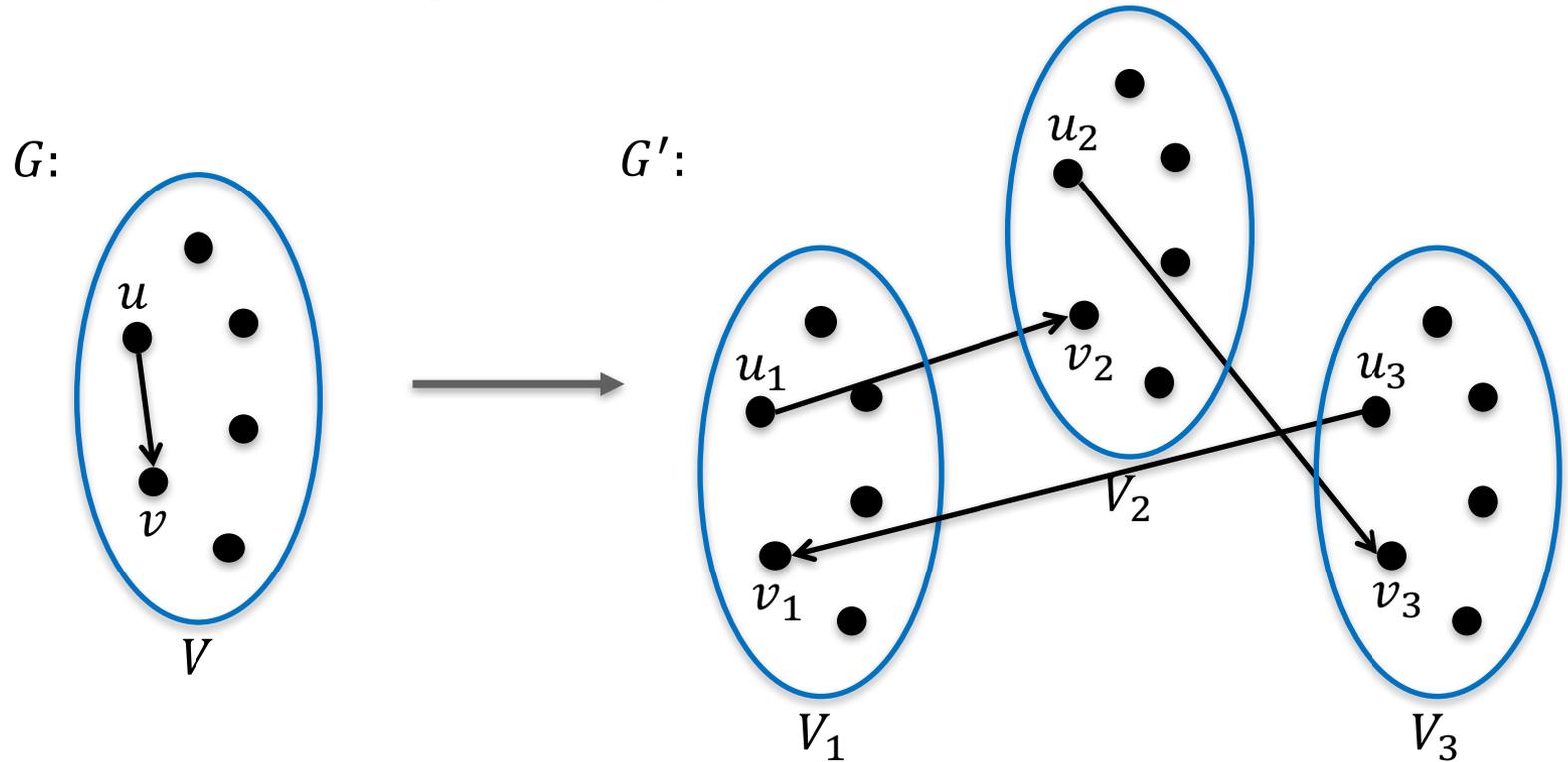
$$w_{j,i} + w_{i,k} + w_{k,j} = w_i + w_j + w_k$$

Node-Weighted Negative Triangle



Node-Weighted Minimum Weight Triangle

we can assume that the graph is **tripartite**:



triangle i, j, k

\Leftrightarrow

triangle i_1, j_2, k_3

Node-Weighted Minimum Weight Triangle

given graph $G = (V, E)$ with $V = I \cup J \cup K$ and node weights w_v ,
compute minimum weight q s.t.

there are $i \in I, j \in J, k \in K$ with $(i, j), (j, k), (k, i) \in E$ and $w_i + w_j + w_k = q$

- assume that I, J, K are sorted by weight
- parameter p (=sufficiently large constant)
- assume $n := |I| = |J| = |K| = p^\ell$ for some $\ell \in \mathbb{N}$
(add isolated dummy vertices)



Node-Weighted Minimum Weight Triangle

given graph $G = (V, E)$ with $V = I \cup J \cup K$ and node weights w_v ,
compute minimum weight q s.t.

there are $i \in I, j \in J, k \in K$ with $(i, j), (j, k), (k, i) \in E$ and $w_i + w_j + w_k = q$

- assume that I, J, K are sorted by weight
- parameter p (=sufficiently large constant)
- assume $n := |I| = |J| = |K| = p^\ell$ for some $\ell \in \mathbb{N}$
(add isolated dummy vertices)

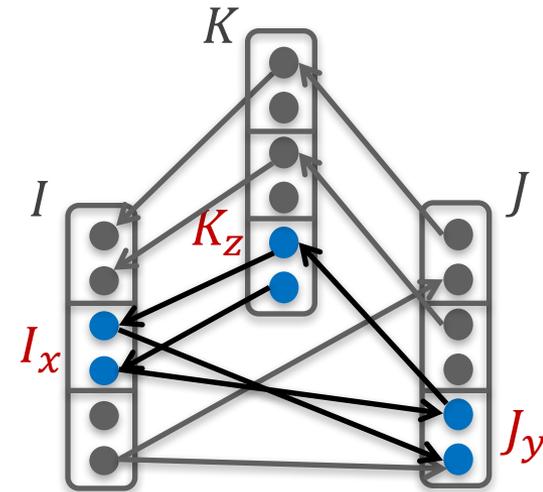
ALG(G): 0) if $n = O(1)$ then solve in constant time

1) split $I = I_1 \cup \dots \cup I_p, J = J_1 \cup \dots \cup J_p, K = K_1 \cup \dots \cup K_p$
(in sorted order: $\max w(I_x) \leq \min w(I_{x+1})$ and so on)

2) $R := \{(x, y, z) \in \{1, \dots, p\}^3 \mid G[I_x \cup J_y \cup K_z] \text{ contains a triangle}\}$

3) for each $(x, y, z) \in R$ s.t. there is no $(x', y', z') \in R$ with $x' < x, y' < y, z' < z$

run $\text{ALG}(G[I_x \cup J_y \cup K_z])$



Node-Weighted Minimum Weight Triangle

3) for each $(x, y, z) \in R$ s.t. there is no $(x', y', z') \in R$ with $x' < x, y' < y, z' < z$
run $\text{ALG}(G[I_x \cup J_y \cup K_z])$

Correctness:

if there is no triangle in $G[I_x \cup J_y \cup K_z]$ then we can ignore it

if $(x, y, z) \in R$ is dominated by $(x', y', z') \in R$:

let i, j, k be a triangle in $G[I_x \cup J_y \cup K_z]$, and i', j', k' a triangle in $G[I_{x'} \cup J_{y'} \cup K_{z'}]$

then $w_i + w_j + w_k \geq \min w(I_x) + \min w(J_y) + \min w(K_z)$

VI

VI

VI

and $w_{i'} + w_{j'} + w_{k'} \leq \max w(I_{x'}) + \max w(J_{y'}) + \max w(K_{z'})$

so we can safely ignore $G[I_x \cup J_y \cup K_z]$



Node-Weighted Minimum Weight Triangle

given graph $G = (V, E)$ with $V = I \cup J \cup K$ and node weights w_v ,
compute minimum weight q s.t.

there are $i \in I, j \in J, k \in K$ with $(i, j), (j, k), (k, i) \in E$ and $w_i + w_j + w_k = q$

- assume that I, J, K are sorted by weight
- parameter p (=sufficiently large constant)
- assume $n := |I| = |J| = |K| = p^\ell$ for some $\ell \in \mathbb{N}$
(add isolated dummy vertices)

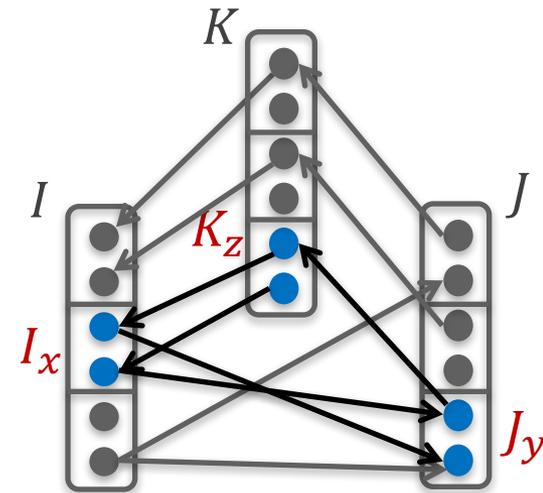
ALG(G): 0) if $n = O(1)$ then solve in constant time

1) split $I = I_1 \cup \dots \cup I_p, J = J_1 \cup \dots \cup J_p, K = K_1 \cup \dots \cup K_p$
(in sorted order: $\max I_x \leq \min I_{x+1}$ aso.)

2) $R := \{(x, y, z) \in \{1, \dots, p\}^3 \mid G[I_x \cup J_y \cup K_z] \text{ contains a triangle}\}$ $O(p^3 n^\omega)$

3) for each $(x, y, z) \in R$ s.t. there is no $(x', y', z') \in R$ with $x' < x, y' < y, z' < z$

run $\text{ALG}(G[I_x \cup J_y \cup K_z])$



↑
how many iterations?
size n/p

Node-Weighted Minimum Weight Triangle

3) for each $(x, y, z) \in R$ s.t. there is no $(x', y', z') \in R$ with $x' < x$, $y' < y$, $z' < z$

How many iterations?

$$\begin{aligned} \text{define } \Delta_{r,s} &:= \{(x, y, z) \in \{1, \dots, p\}^3 \mid x - y = r \text{ and } x - z = s\} \\ &= \{(1, 1 - r, 1 - s), (2, 2 - r, 2 - s), \dots, (p, p - r, p - s)\} \cap \{1, \dots, p\}^3 \end{aligned}$$

for $r, s \in \{-p, \dots, p\}$

the sets $\Delta_{r,s}$ cover $\{1, \dots, p\}^3$

line 3) applies to at most one (x, y, z) in $\Delta_{r,s}$ for any r, s !

hence, there are at most $(2p)^2 = 4p^2$ recursive calls to ALG



Node-Weighted Minimum Weight Triangle

given graph $G = (V, E)$ with $V = I \cup J \cup K$ and node weights w_v ,
compute minimum weight q s.t.

there are $i \in I, j \in J, k \in K$ with $(i, j), (j, k), (k, i) \in E$ and $w_i + w_j + w_k = q$

- assume that I, J, K are sorted by weight
- parameter p (=sufficiently large constant)
- assume $n := |I| = |J| = |K| = p^\ell$ for some $\ell \in \mathbb{N}$
(add isolated dummy vertices)

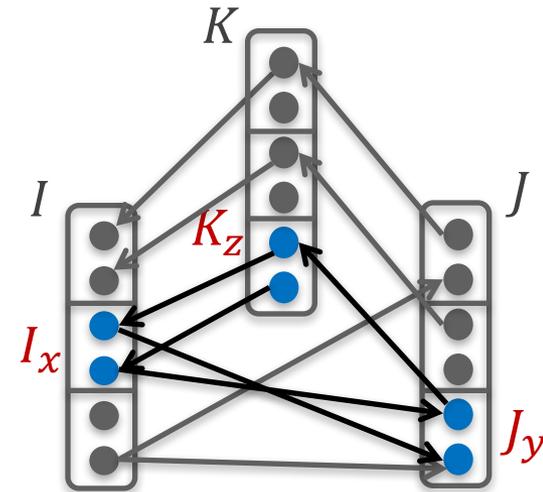
ALG(G): 0) if $n = O(1)$ then solve in constant time

1) split $I = I_1 \cup \dots \cup I_p, J = J_1 \cup \dots \cup J_p, K = K_1 \cup \dots \cup K_p$
(in sorted order: $\max I_x \leq \min I_{x+1}$ aso.)

2) $R := \{(x, y, z) \in \{1, \dots, p\}^3 \mid G[I_x \cup J_y \cup K_z] \text{ contains a triangle}\}$ $O(p^3 n^\omega)$

3) for each $(x, y, z) \in R$ s.t. there is no $(x', y', z') \in R$ with $x' < x, y' < y, z' < z$

run $\text{ALG}(G[I_x \cup J_y \cup K_z])$ size n/p



Node-Weighted Minimum Weight Triangle

recursion: $T(n) \leq 4p^2 \cdot T(n/p) + O(p^3 n^\omega)$

$$T(n) \leq 4p^2 \cdot T(n/p) + \alpha p^3 n^\omega \quad \text{for some constant } \alpha$$

want to show: $T(n) \leq 2\alpha p^3 n^\omega$

plug in:
$$\begin{aligned} T(n) &\leq 4p^2 \cdot 2\alpha p^3 (n/p)^\omega + \alpha p^3 n^\omega \\ &= \alpha p^3 n^\omega (1 + 8p^{2-\omega}) \end{aligned}$$

assume $\omega > 2$: $\leq 2\alpha p^3 n^\omega$ for a sufficiently large constant p

so we have: $T(n) \leq O(n^\omega)$

(if $\omega = 2$: show that $T(n) \leq 2\alpha p^3 n^{\omega+\varepsilon}$)

in total: $T(n) \leq O(n^\omega + n^{2+\varepsilon})$ for any $\varepsilon > 0$



In this lecture you learned that...

...(B)MM is useful for **designing theoretically fast algorithms**

- Exercise: k-Clique in $O(n^{\omega k/3})$
- Exercise: MaxCut in $O(2^{\omega n/3} \text{poly}(n))$
- Node-Weighted Negative Triangle in $O(n^\omega)$

...BMM is an **obstacle for practically fast / theoretically very fast algorithms**

no combinatorial $O(n^{3-\epsilon})$

not faster than $O(n^{2.373})$

- Transitive Closure has no $O(n^{\omega-\epsilon})$ / combinatorial $O(n^{3-\epsilon})$ algorithm
- Exercise: pattern matching with 2 patterns
- Sliding Window Hamming Distance
- context-free grammar problems

