## Weighted Min-Cut: A Cross-Paradigm Algorithm

Sagnik Mukhopadhyay

University of Sheffield, UK

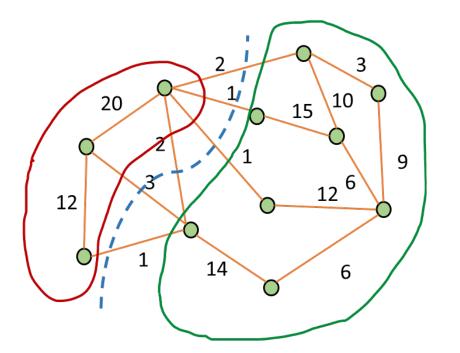


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### Weighted min-cut and our results

## The (global) mincut problem

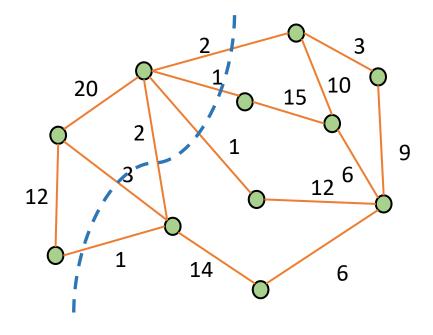
Remove edges to disconnect the graph (minimize total weight)



"[This problem] plays an important role in the <u>design of</u> <u>communication networks</u>. If a few links are cut ..."

LEDA (<u>http://www.algorithmic-solutions.info/</u>)

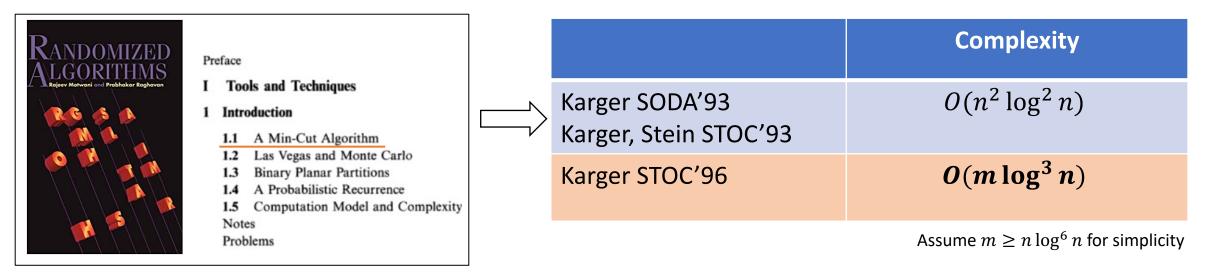
## The (global) mincut problem



- Cut:
  - Set of edges whose removal disconnects *G*.
- Min-cut:
  - Cut with minimum total edge weight.

**Goal:** Find a min-cut in a weighted graph.

## State of the arts: near-linear time



#### Problem:

# **Complicated dynamic programming & data structures**

# Hard to adapt to new computational models

No efficient algorithms in distributed, streaming, query complexity

n=#nodes, m=#edges

### Models of computation

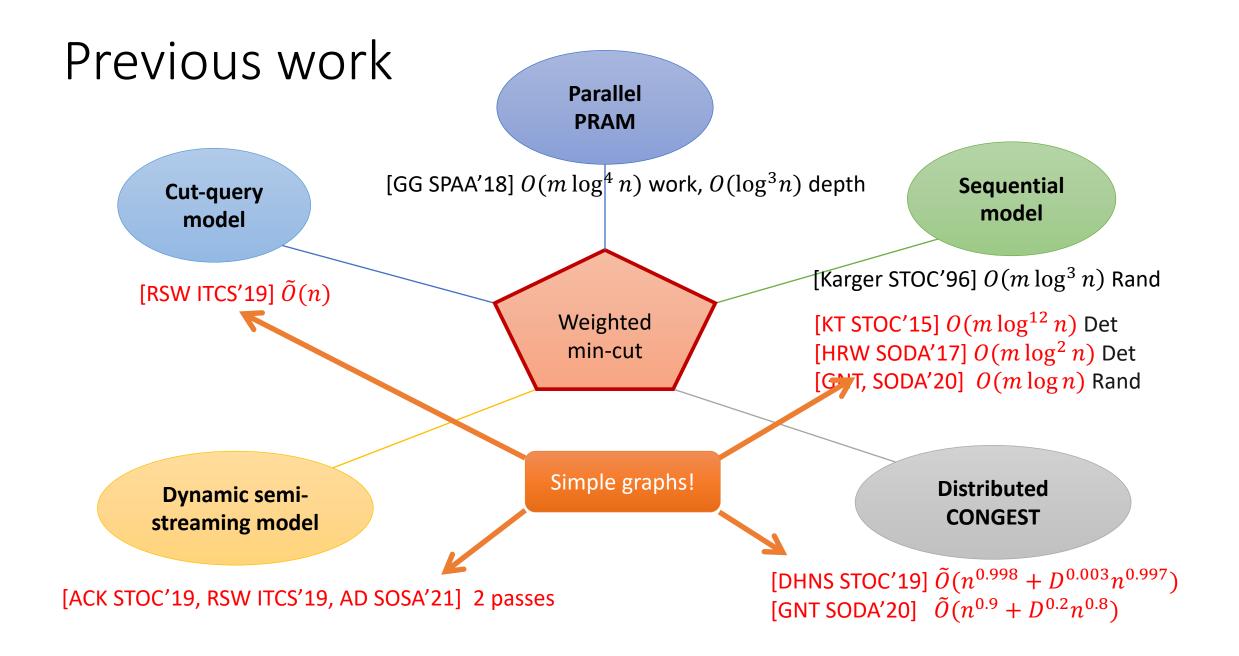
**Cut Query:** Allowed to query arbitrary cuts of *G*. Charged once per query.

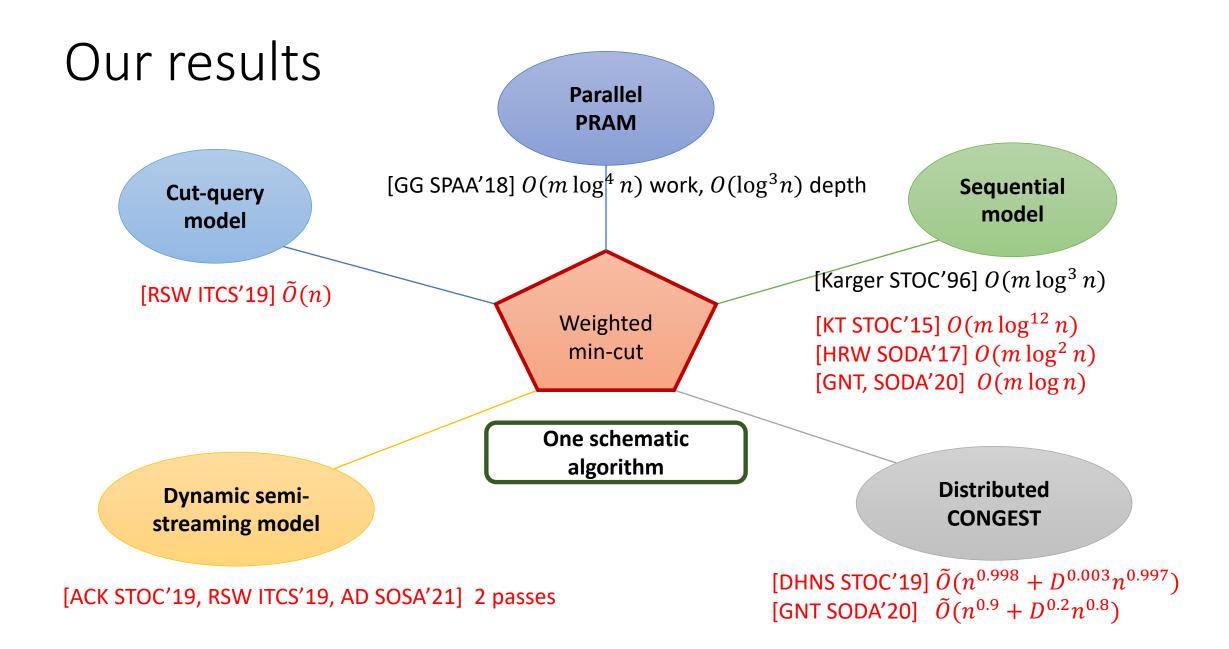
**Dynamic semi-streaming:**  $O(n \operatorname{polylog} n)$  bits of internal memory. Charged once per pass.

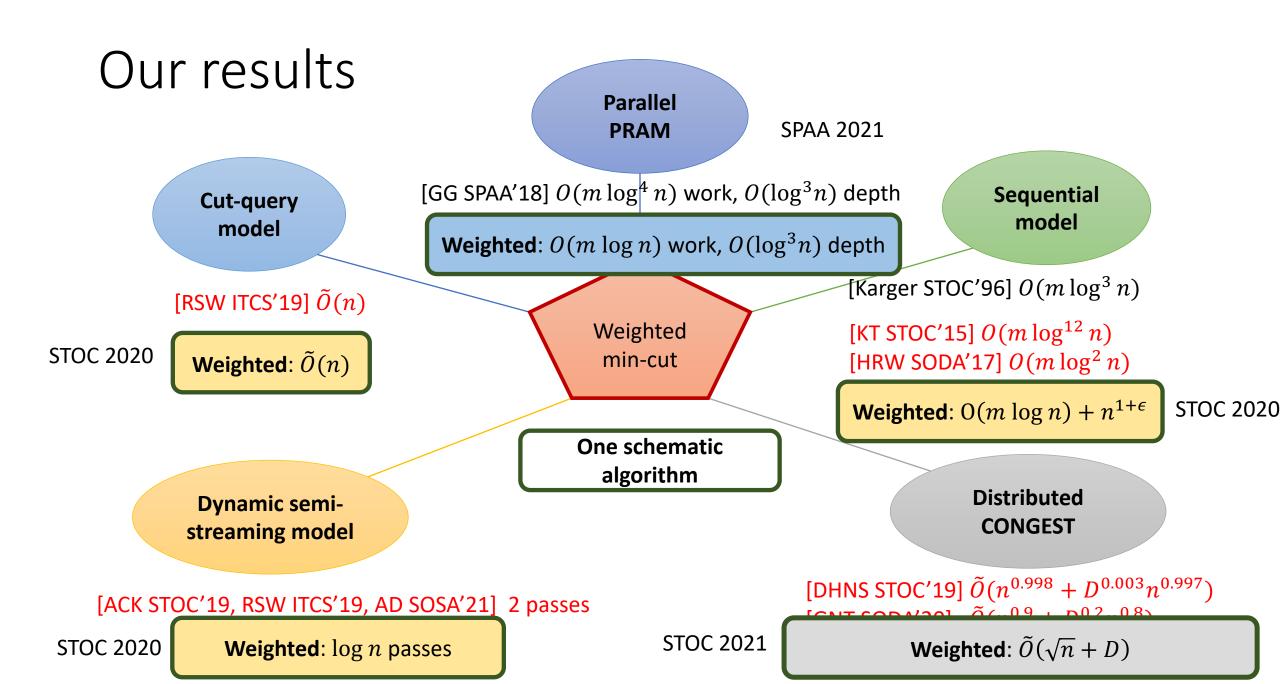
Sequential: Standard unit-cost RAM model.

Parallel PRAM: Concurrent read exclusive write. Complexity is (work, depth) of computation.

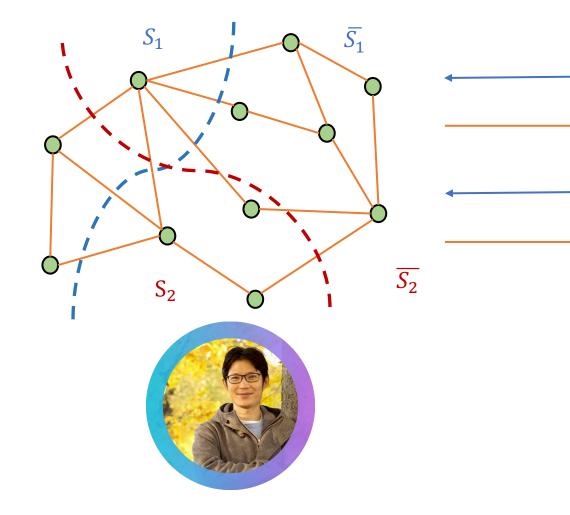
**Distributed CONGEST:** Bandwidth restricted ( $O(\log n)$  bits per round). Charged once per round.







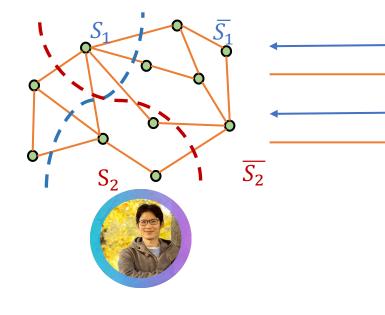
## Today: Cut-query model



A cut  $(S_1, \overline{S_1})$ Value of the cut  $(S_1, \overline{S_1})$ A cut  $(S_2, \overline{S_2})$ Value of the cut  $(S_2, \overline{S_2})$ 



## Today: Cut-query model



A cut 
$$(S_1, \overline{S_1})$$
  
Value of the cut  $(S_1, \overline{S_1})$   
A cut  $(S_2, \overline{S_2})$   
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Graph cuts ⇔ Submodular function

Cut-query ⇔ Submodular function minimization via query

## Our result

**Common barrier**: How to overcome?

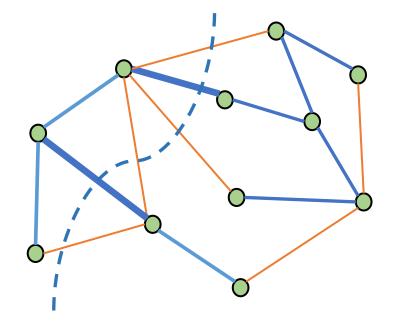
#### Remark:

- Assuming simple graph makes life a lot easier!
- None of the improvements on simple graphs follow Karger's framework.

What is so hard about Karger's framework?

# 2-respecting cut

• Main subroutine of Karger's algorithm.



- Spanning tree:
  - A tree T with edges from E(G) that spans all vertices.
- 2-respecting cut:
  - Cut C with at most 2 edges from T.

# 2-respecting cut

- 2-respecting cut: • Cut C with at most 2 edges from T. **Not 2-respecting**
- Spanning tree:
  - A tree T with edges from E(G) that spans all vertices.

# 2-respecting cut

- Spanning tree:
  - A tree *T* with edges from E(*G*) that spans all vertices.
- 2-respecting cut:
  - Cut C with at most

Main bottleneck!

**Goal:** Find a 2 –respecting min-cut in a weighted graph given a spanning tree *T*.

**Theorem (Karger).** Efficient algorithm for solving 2-respecting min-cut implies efficient algorithm for solving min-cut.

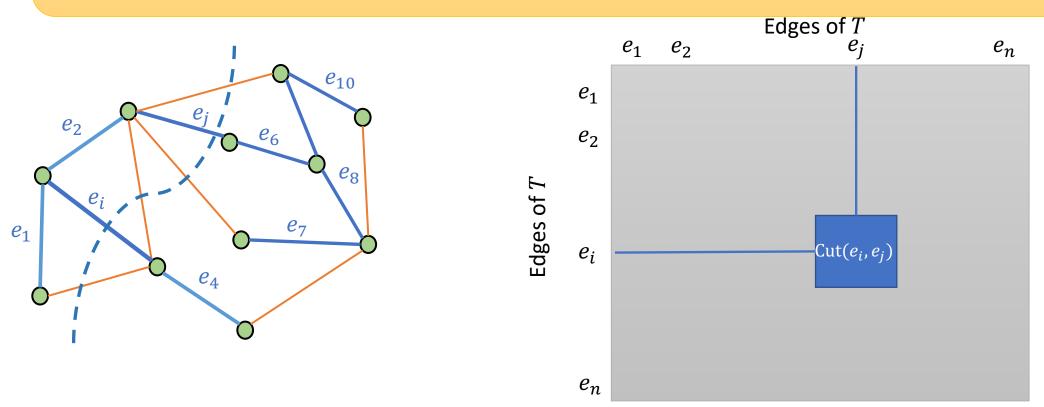
### Our result

### New improved algorithm for min 2-resp cut problem.

### One (schematic) algorithm that works across models

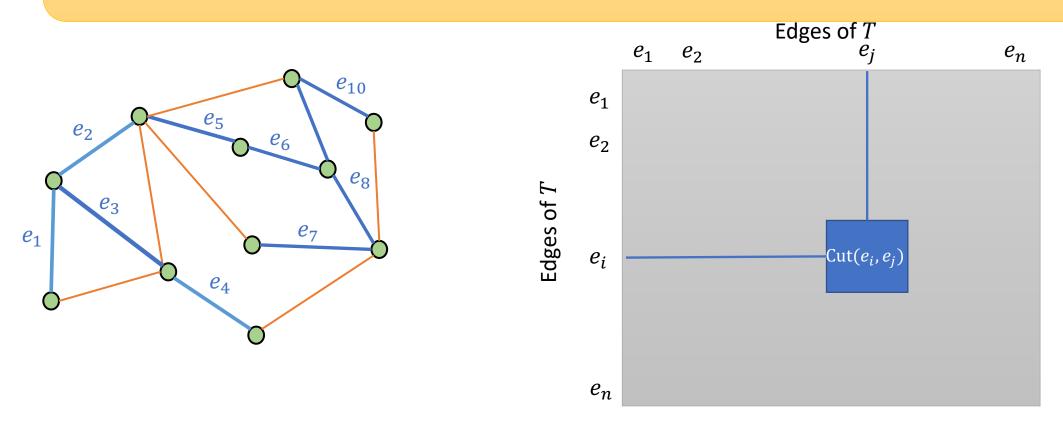
### A closer look...

# Simple and efficient algorithm for **min 2-resp cut** for all models

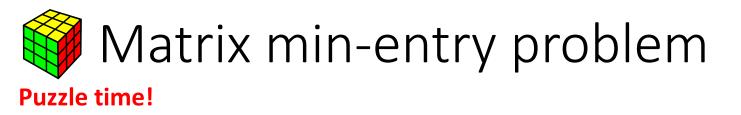


### A closer look...

- $n^2$  many entries to look up.
- Goal: Minimize the number of look-ups to find a minimum value.

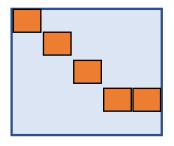


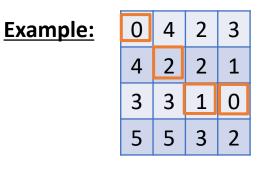
#### A quick detour: Matrix min-entry puzzle



<u>Input:</u> I will write down an  $n \times n$  matrix M (you don't see M but know n).

Promise: Monotonicity - column minimums never move up as we go right

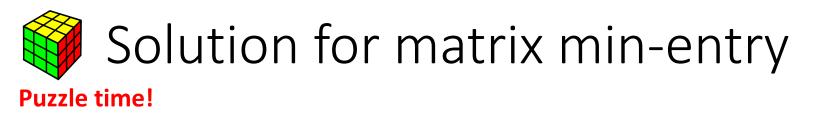




= min of each column

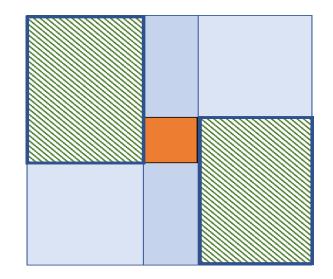
<u>Goal:</u> Find the minimum entry in M.

<u>Query</u>: You can ask for one entry at a time. How many entries do you need?



 $O(n \log n)$  queries by divide and conquer.

- 1. Find min in the middle column
  - by asking for everything in that column.
- 2. The recurse on both sides.
- 3. Search area decreases by half.



In fact, there is an O(n)-query solution [SMAWK]. But it's not useful for us (too sequential)

### Matrix min-entry to Global min-cut

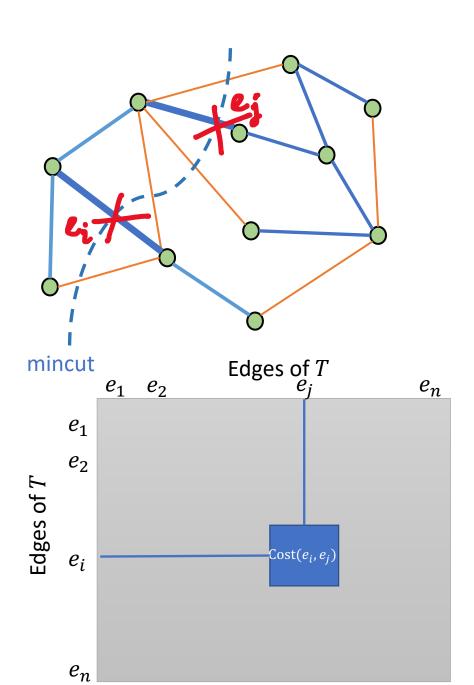
# $Q_{\text{Technical}}$ Mincut $\rightarrow$ Matrix min-entry

2-respecting min-cut: Easy to find spanning tree **T** that 2constains the min-cut.

**Cost matrix**: Guess two tree-edges crossing the cuts Entries in  $n \times n$  matrix There are  $O(n^2)$  candidate

cuts

Cut-query model: Can find **1** entry by **1** query.



# $Q_{\text{Technical}}$ Mincut $\rightarrow$ Matrix min-entry

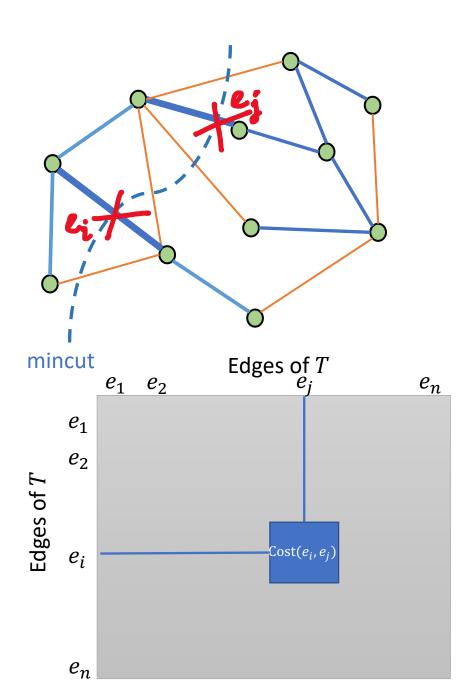
2-respecting min-cut: Easy to find spanning tree **T** that 2constains the min-cut.

**Cost matrix**: Guess two tree-edges crossing the cuts Entries in  $n \times n$  matrix

There are  $O(n^2)$  candidate cuts

Too expensive to compute all matrix entries

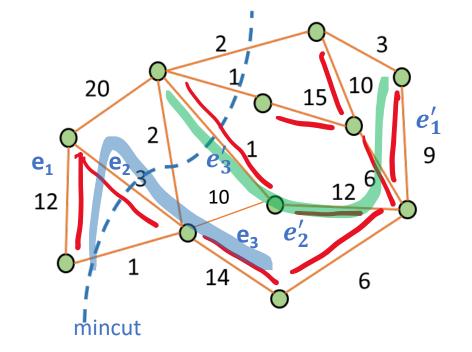
1 stream-pass can compute only **n** entries.

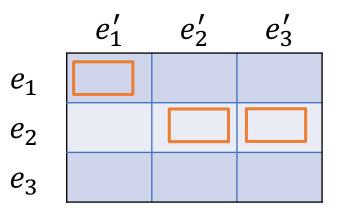


# $Q_{\text{Technical}}$ Mincut $\rightarrow$ Matrix min-entry

**<u>Cute trick</u>: Assume** we know two paths in **T** where the best candidates are in

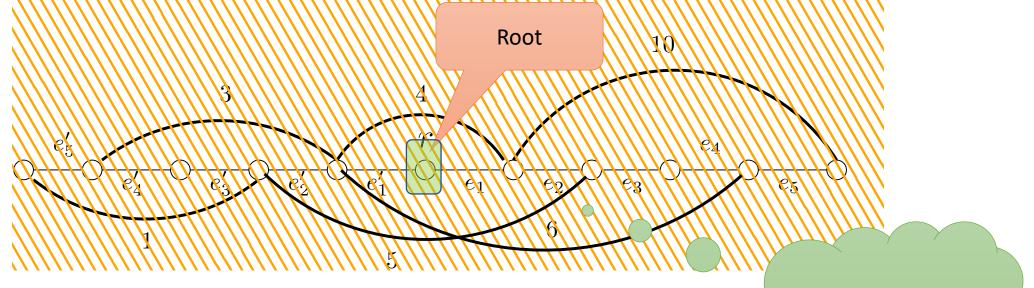
**Lemma:** The corresponding cost matrix is **monotone** like in the puzzle!





Monotone matrix

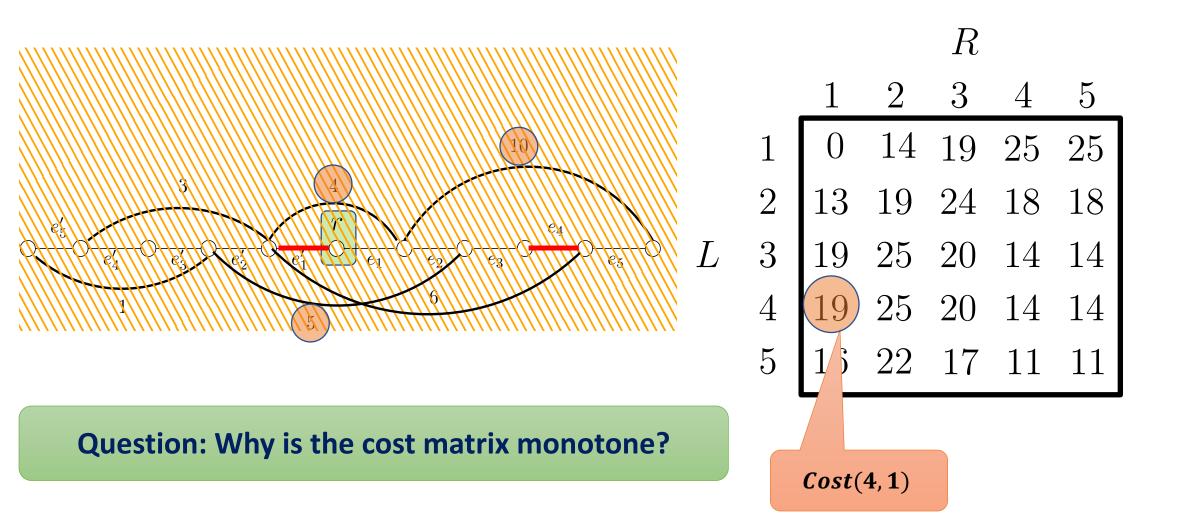
# Key idea: Spanning tree with 2 paths

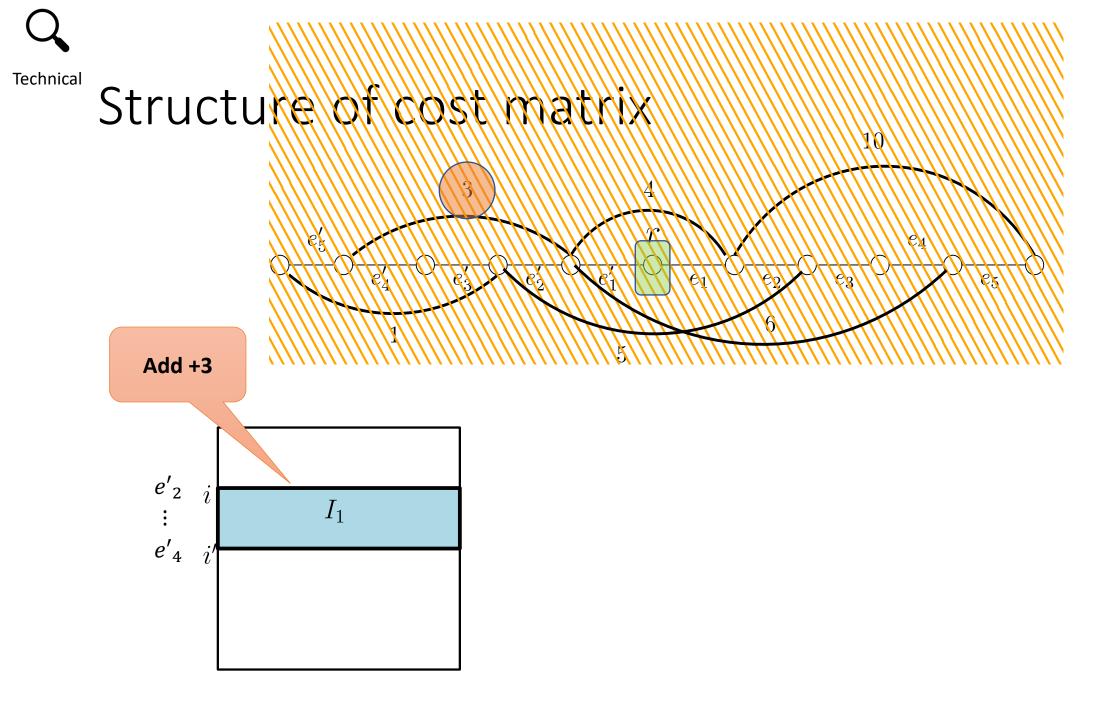


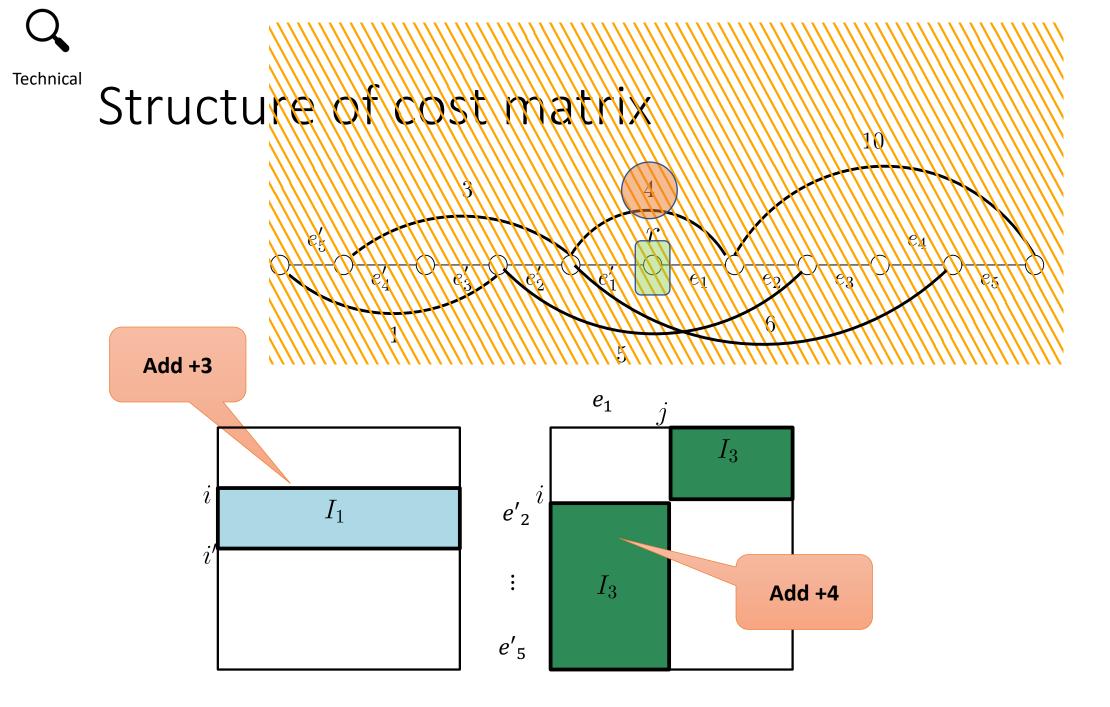
Bipartite path problem

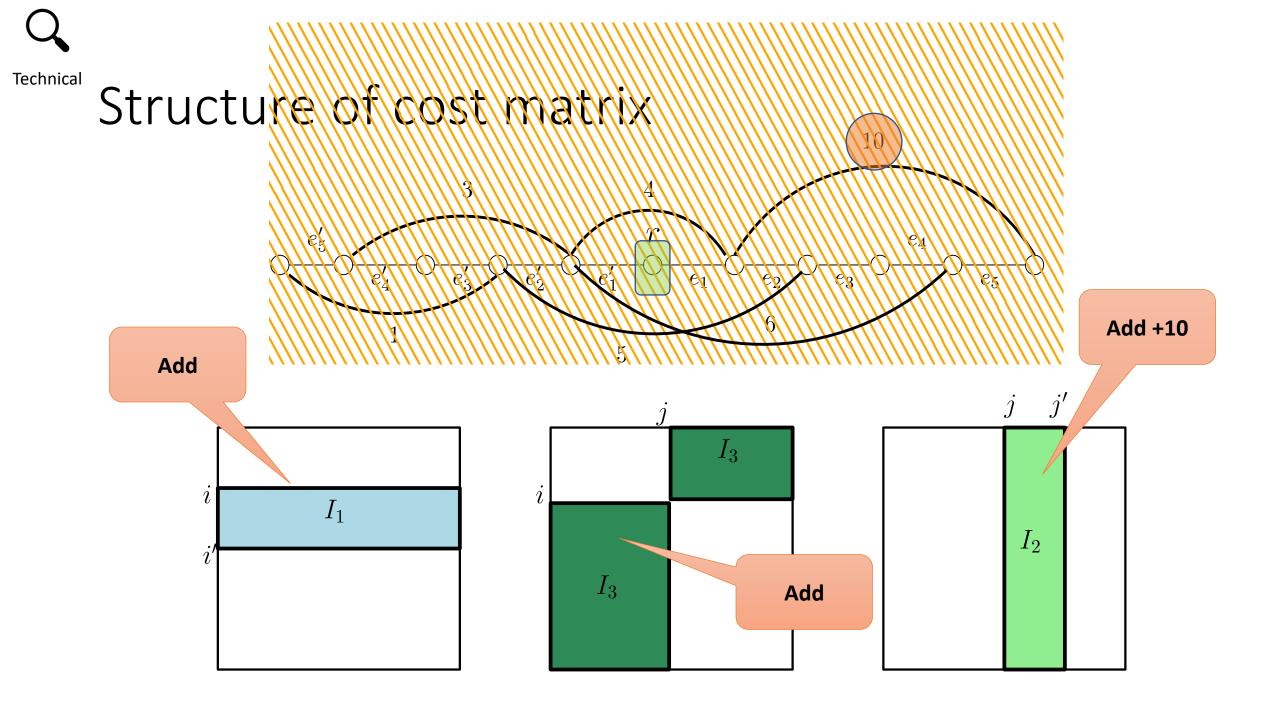
 $Cost(e_i, e_j), e_i \in L, e_j \in R$ 



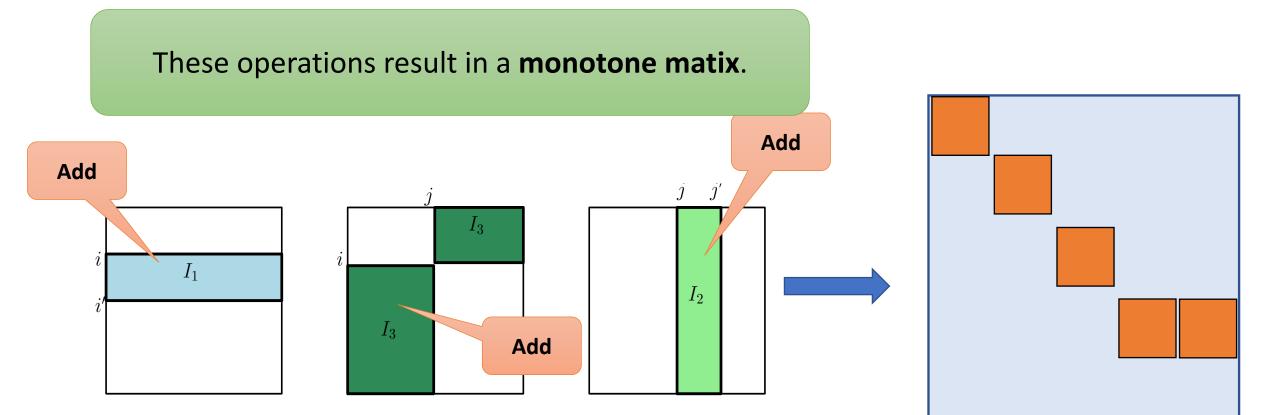








# Technical Monotonicity of cost matrix



• Queries needed to solve bipartite path problem =  $O(n \log n)$ .

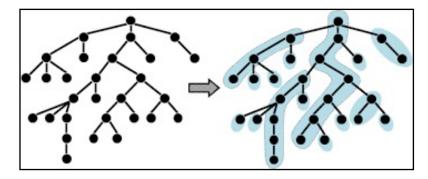
# $\mathbf{Q}_{\text{Technical}} \text{Mincut} \rightarrow \text{Matrix min-entry (2)}$

**<u>Cute trick</u>: Assume** we know two paths in **T** where the best candidates are in

**Lemma:** The corresponding cost matrix is **monotone** like in the puzzle!

Decompose **T** into paths using **path decomposition** (e.g. heavy-light, bough/layering decomposition) & the simulate matrix min-entry algorithm for **some** pairs of paths

Leftover details: Picking a few pairs

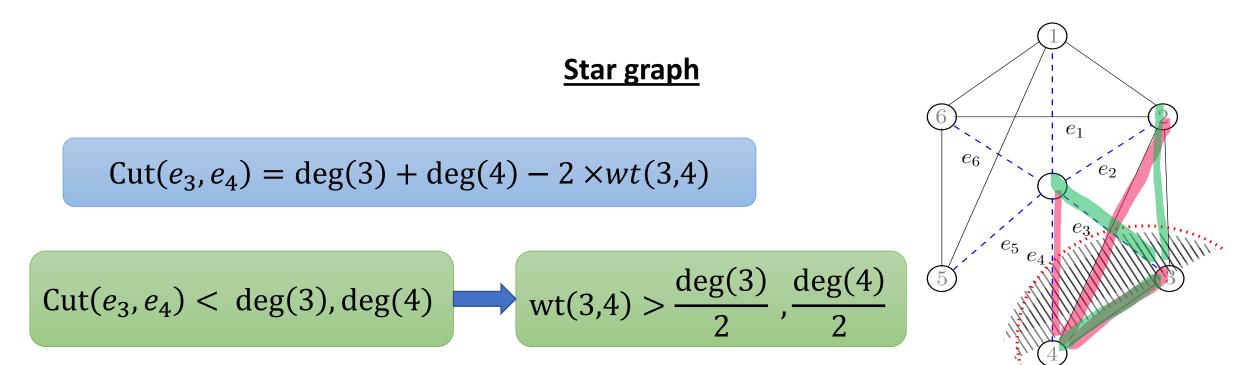


Heavy-light decomposition

#### Choosing a few pairs from decomposition

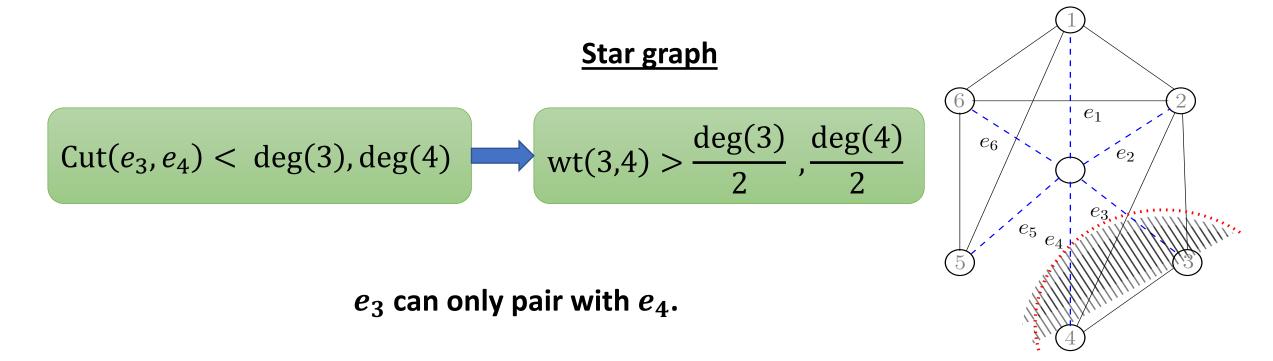
#### Picking up few pairs of path

Lemma (Not quite true): Each path can be paired with only one path such that the minimum 2-respect cut belongs to one such pair.



#### Picking up few pairs of path

**Lemma (Not quite true):** Each path can be paired with **only one path** such that the minimum 2-respect cut belongs to one such pair.



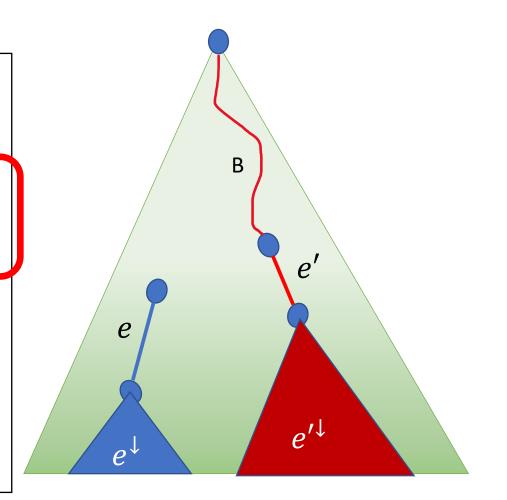
Lemma (Not quite true): Each path can be paired with only one path such that the minimum 2-respect cut belongs to one such pair.

Lemma (Almost true): Each path can be paired with only one root-to-leaf path such that the minimum 2-respect cut belongs to one such pair.

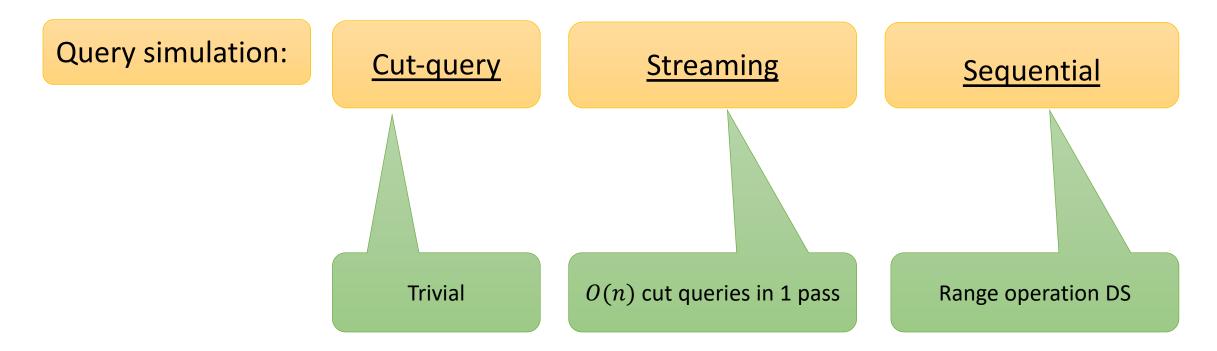
log *n* many paths from the tree decomposition

# All results follow from one schematic algorithm (with different implementation details)

- 1. Decompose **T** into paths using bough/layering decomposition.
- 2. Each edge **e** in **A** picks path **B** that is interesting to it.
- 3. Simulate the matrix min-entry algorithm on the cost matrix of every such pair (A,B) after contracting useless edges.



## Cut-query simulation



More on range DS

## Summary

- Spanning tree packing. Cut sparsifier with random neighbor sampling
- 2-respecting min-cut is the main bottle-neck of min-cut.
- Spanning tree: Path
  - Monotone matrix

Can be solved quickly using properties of

- Spanning tree: Star-graph Can be solved quickly using *edge pairing*
- Putting it all together: General spanning tree

Tree decomposition and **almost correct lemma** 

Skipped this!

Karger framework

### Open problems

## Open problems

- Graph problems in 2-party communication setting.
- Min-cut in dynamic setting.
- Directed min-cut and vertex connectivity.
- Other graph problems admitting cross-paradigm algorithm (?)

## Thank you.

#### Simulating a cut-query

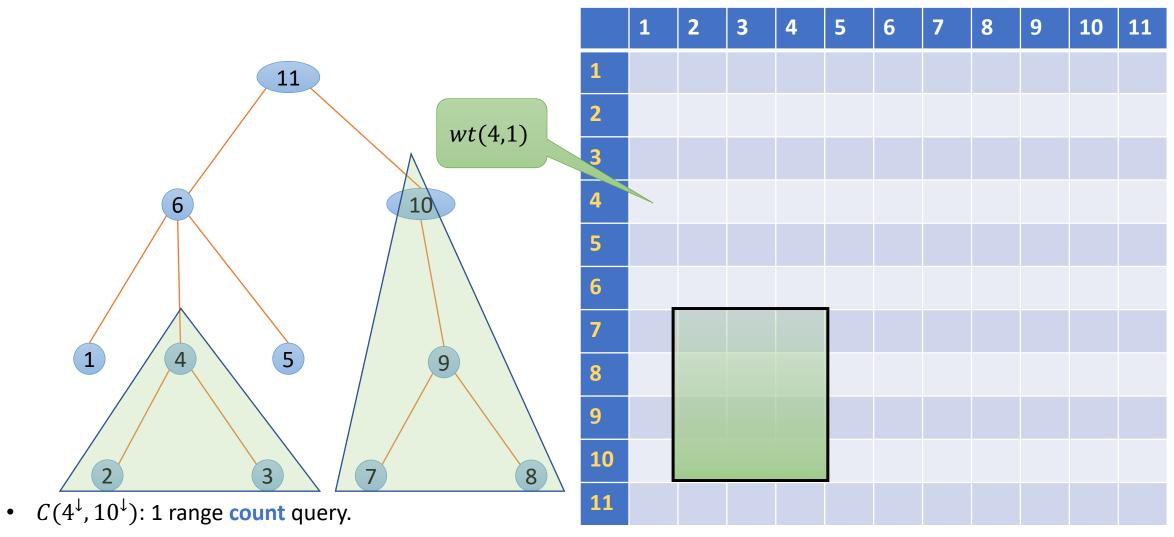
### Range operation data-structure

arity x depth

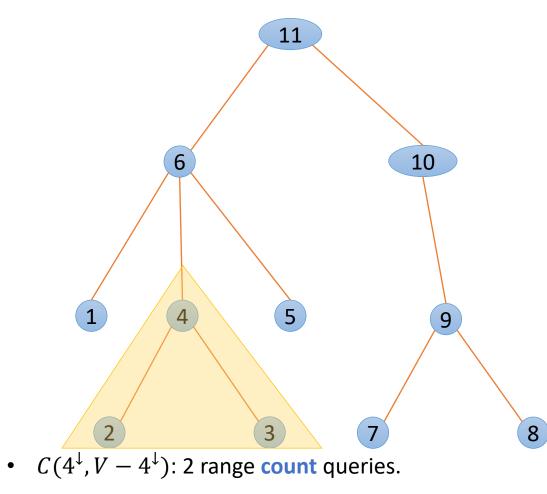
- Takes *m* points in a 2-d plane.
- Preprocessing: O(m) # elements x depth
- Query: An axis-aligned rectangle *R*.
  - Count points in  $R: O(n^{\epsilon})$ .
  - Sample point from  $R: O(n^{\epsilon})$ .
    - Amortized.

R 

### Range operation on spanning tree

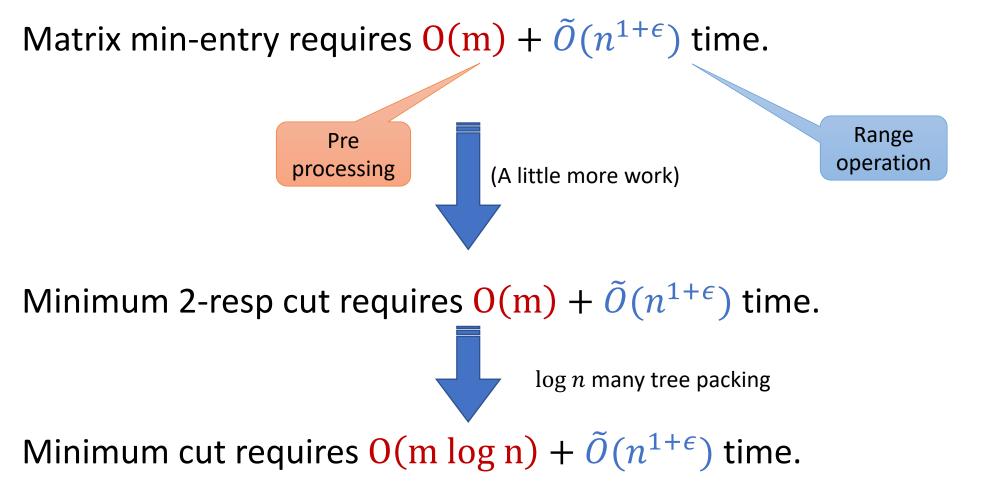


### Range operation on spanning tree



	1	2	3	4	5	6	7	8	9	10	11
1											
2											
3											
4											
5											
6											
7											
8											
9											
10											
11											

### Range operation on spanning tree



Back to summary