

How to go from partial to full retroactivity in detail

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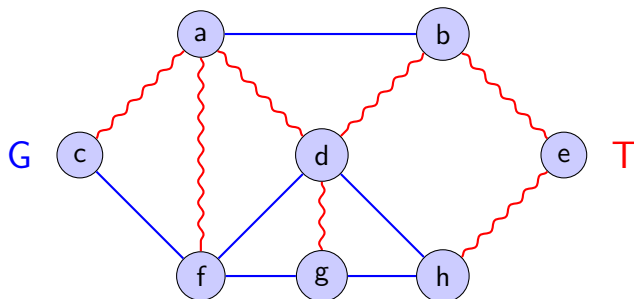


Figure: Graph G (blue edges) and spanning tree T (red wavy edges)

Minimum Spanning Tree and Forest

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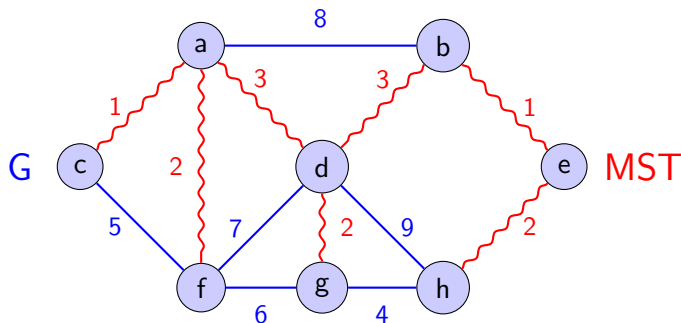


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- **Problem:** Keep track of an MSF in a graph that grows over time
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- **Solution:** Frederickson (1983) using link-cut trees

Frederickson's link-cut tree solution

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 - ▶ $\text{find_max}(u, v)$: $\mathcal{O}(\log n)$ amortized
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- **Algorithm for adding edge (u, v, w) :**
 - 1 Check if u and v are in same component
 - 2 If not: add edge to forest
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 - 4 If $w < \text{max cost}$: replace max edge with new edge

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- **Total cost:** Amortized $\mathcal{O}(\log n)$ per edge addition

Incremental MSF example - Step 1

- **add_edge(g, h, 4):** Add edge with cost 4

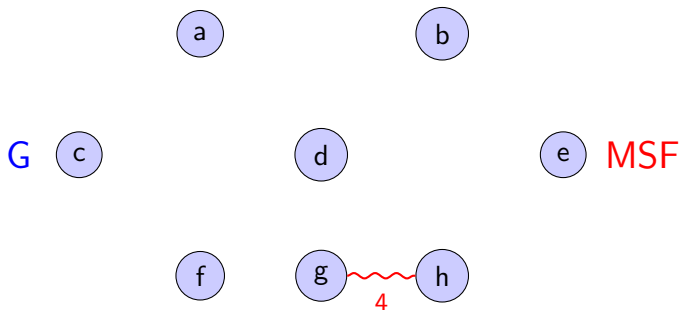


Figure: Step 1: Added edge (g,h) with cost 4

- **MSF:** $\{g-h\}$

Incremental MSF example - Step 2

- **add_edge(c, a, 1):** Add edge with cost 1

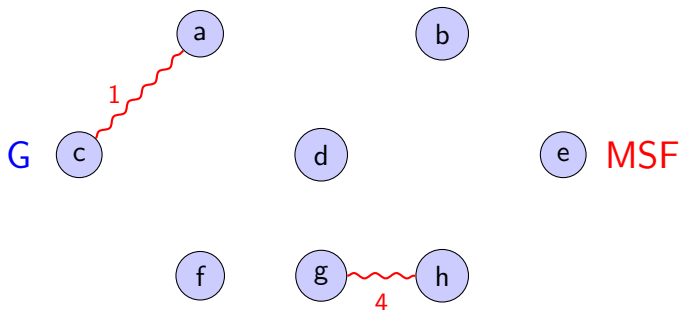


Figure: Step 2: Added edge (c,a) with cost 1

- **MSF:** $\{g-h, c-a\}$

Incremental MSF example - Step 3

- **add_edge(f, g, 6):** Add edge with cost 6

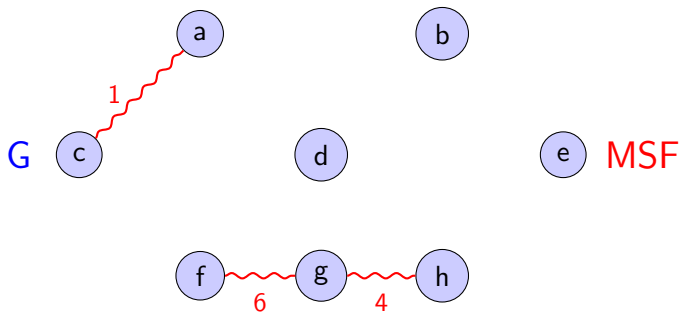


Figure: Step 3: Added edge (f,g) with cost 6

- **MSF:** $\{g-h, c-a, f-g\}$

Incremental MSF example - Step 4

- **add_edge(a, f, 2):** Add edge with cost 2

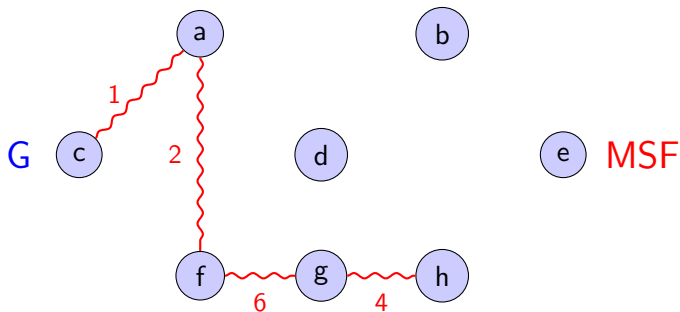


Figure: Step 4: Added edge (a,f) with cost 2

- **MSF:** $\{g-h, c-a, f-g, a-f\}$

Incremental MSF example - Step 5

- **add_edge(c, f, 5):** Add edge with cost 5

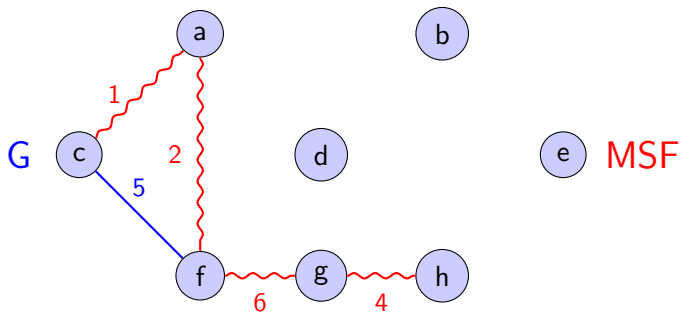


Figure: Step 5: Added edge (c,f) with cost 5

- **MSF:** $\{g-h, c-a, f-g, a-f\}$

Incremental MSF example - Step 6

- **add_edge(f, d, 7):** Add edge with cost 7

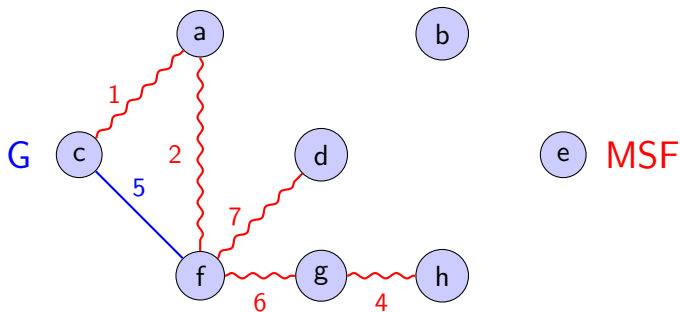


Figure: Step 6: Added edge (f,d) with cost 7

- **MSF:** $\{g-h, c-a, f-g, a-f, f-d\}$

Incremental MSF example - Step 7

- **add_edge(a, d, 3):** Add edge with cost 3

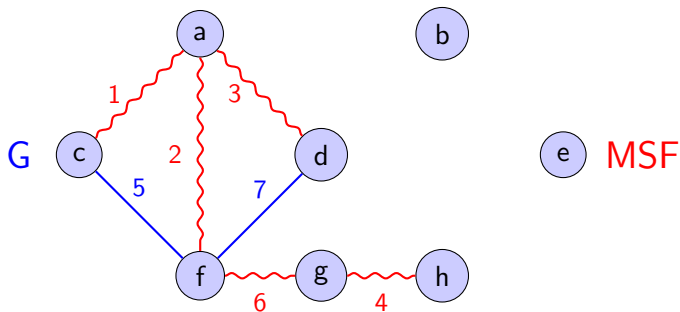


Figure: Step 7: Added edge (a,d) with cost 3

- **MSF:** $\{g-h, c-a, f-g, a-f, a-d\}$

Incremental MSF example - Step 8

- **add_edge(d, g, 2):** Add edge with cost 2

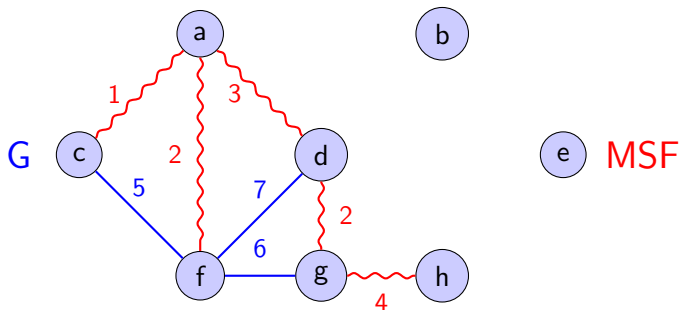


Figure: Step 8: Added edge (d,g) with cost 2

- **MSF:** $\{g-h, c-a, a-f, a-d, d-g\}$

Incremental MSF example - Final Result

- **Continue adding edges...**

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- **Final MSF:** Minimum spanning forest with optimal cost

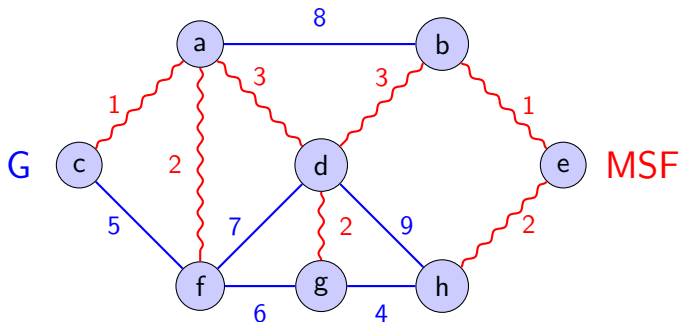


Figure: Final MSF with optimal cost = 14

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- **Retroactivity:** Manipulate the sequence of updates
- **Operations:**
 - ▶ Insert update at time t (possibly in the past)
 - ▶ Remove update at time t
 - ▶ Query at time t (not just present)

Partial vs Full retroactivity

Fully Retroactive

- Queries at **any** time t
- Insert/remove updates at any time

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Semi-Retroactive

- Queries at **any** time t
- Insert updates at any time
- **No removal** of updates

The challenge

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How to transform partial \rightarrow full retroactivity?

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- **Problem:** Need to support queries at any time t
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- **Solution approach:** Square-root decomposition
- **Key insight:** Keep checkpoints with data structure states
- **Implementation:** Demaine, Iacono & Langerman (2007)

Demaine, Iacono & Langerman's solution

Theorem (Theorem 05)

Any partially retroactive data structure can be transformed into a fully retroactive one with:

- $\mathcal{O}(\sqrt{m})$ slowdown per operation
- $\mathcal{O}(m)$ space usage
- **Requirement:** Need persistent version of the data structure

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- **Key idea:** Square-root decomposition
- Keep \sqrt{m} checkpoints with data structure states
- **Query at time t :**
 - 1 Find closest checkpoint before t
 - 2 Apply updates from checkpoint to t
 - 3 Answer query, then rollback

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What if we don't have or don't want to use persistent data structures?

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Our contribution

Simple rebuilding strategy without persistent data structures

- Same time complexity: $\mathcal{O}(\sqrt{m})$ per operation
- Space usage: $\Theta(m\sqrt{m})$

Starting point

- **Junior & Seabra's solution:** Semi-retroactive incremental MSF
- **Operations:**
 - ▶ `add_edge(u, v, w, t)`: add edge at time t
 - ▶ `get_msf(t)`: get MSF at time t

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- **Checkpoints:** $t_i = i\sqrt{m}$ for $i = 1, \dots, \sqrt{m}$

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- **Implementation:** Square-root decomposition
- **Checkpoints:** $t_i = i\sqrt{m}$ for $i = 1, \dots, \sqrt{m}$
- **Data structures:** D_i contains edges before time t_i
- **Time:** $\mathcal{O}(\sqrt{m} \log n)$ per operation

Limitations → Key Insight

Problems with the Existing Static Approach

- **Fixed m :** Requires knowing the maximum sequence length (m) beforehand. Meaning that it cannot handle arbitrary growth or dynamic operation counts.

Our Dynamic Goal and Solution

Goal: Remove the **Fixed m** dependency while preserving time complexity.

- **Key Insight:** Introduce a ****dynamic rebuilding process**** to handle arbitrary growth.
- **Challenge:** Rebuilding \sqrt{m} checkpoints must be fast, avoiding complex persistent data structures.
- **Solution:** ****Reuse**** the existing data structures to efficiently reconstruct new checkpoints.

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- **Time per rebuilding:** $\mathcal{O}(m \log n)$
- **Amortized cost:** $\mathcal{O}(\sqrt{m} \log n)$ per operation

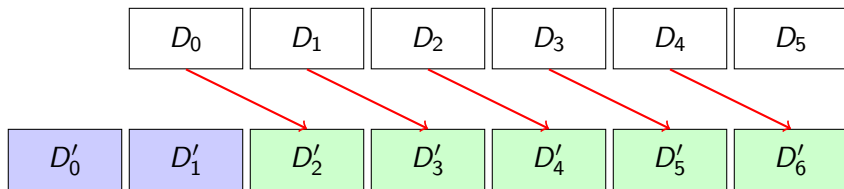
Rebuilding algorithm

- ① $D'_0 \leftarrow \text{NEWINCREMENTALMSF}()$
- ② $D'_1 \leftarrow \text{NEWINCREMENTALMSF}()$
- ③ For $i = 2$ to $k + 1$: $D'_i \leftarrow D_{i-2}$ ▷ reuse existing
- ④ For $i = 1$ to $k + 1$:
 - ▶ $p \leftarrow \text{KTH}(S, i(k + 1))$ ▷ $i(k + 1)$ th edge
 - ▶ $t'_i \leftarrow p.\text{time}$
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- ⑤ Return $k + 1, D', t'$

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Original



$$D_i \rightarrow D'_{i+2}$$

Results

Our contribution

- **General transformation:** Partial \rightarrow Full retroactivity
- **No persistent data structures needed**
- **Same time complexity:** $\mathcal{O}(\sqrt{m})$ per operation
- **Space trade-off:** $\Theta(m\sqrt{m})$ vs $\mathcal{O}(m)$

Results

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Semi-retroactive MSF implementation

- **Operations:** `add_edge(u, v, w, t)`, `get_msf(t)`
- **Time:** $\mathcal{O}(\sqrt{m} \log n)$ per operation
- **Space:** $\Theta(m\sqrt{m})$
- **No fixed m or time range restrictions**

Thank you!

Questions?