Greedy

What is Greedy?

Follows the "looks good" strategy.

Recap the Graph Algorithm

- DFS (walking in a maze)
 - If we can explore, then explore.
 - If we can not explore, backtrack.
 - Do not re-visit a vertex.
 - Applications
 - Cycle
 - Topological
 - SCC

Recap the Graph Algorithm

- BFS (waterfront)
 - 1 step from r
 - 2 steps from r
 - ...
 - Application
 - Shortest Path

Recap the Graph Algorithm

- Dijkstra (a generalized BFS)
 - Explore s.
 - Explore the closet vertex from s.
 - Explore the second closest vertex from s.
 - ...
 - We can use Fibonacci heap to improve it.
- Bellman-Ford

Are they Greedy?

Do we have any other Greedy?

Examples

- Finding Shortest Path
 - Dijkstra.
- Finishing homework
 - Keep finishing the one with the **closest** deadline.

Is that optimal?

Formalize the problem

- Input: n homework, each homework j has a size s_j , and a deadline d_j .
- Output: output a time schedule of doing homework!

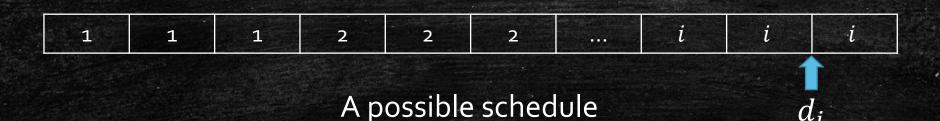
Algorithm

- Greedy
 - Keep finishing the homework with the **closest** deadline.
- Prove it is optimal.
- What is optimal?
- Claim: If we can not finish all the homework by the greedy order, then no one can finish all the homework on time.

Discussion

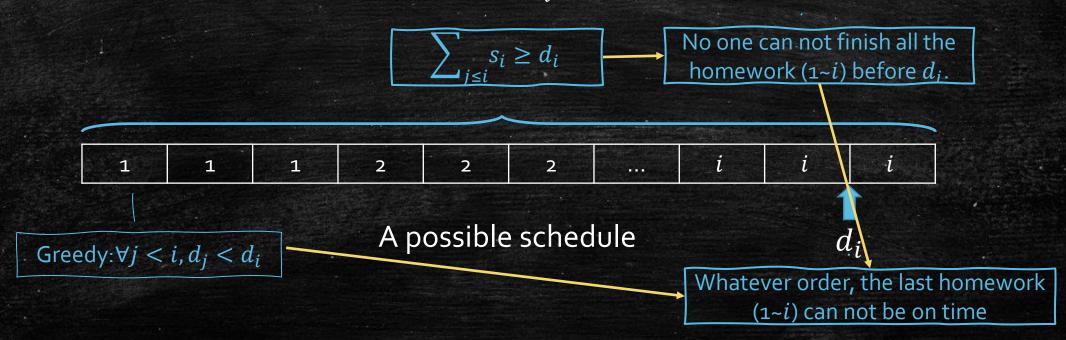
Proof

- Claim: If we can not finish all the homework by the greedy order, then no one can finish all the homework on time.
- Proof:
 - If there exist i, finished later than d_i , what do we have?



Proof

- Claim: If we can not finish all the homework by the greedy order, then no one can finish all the homework on time.
- Proof:
 - If there exist i, finished later than d_i , what do we have?



Minimum Spanning Tree

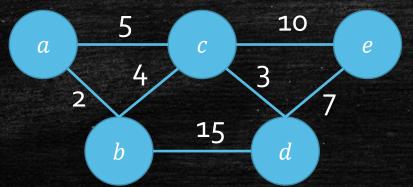
Prime & Kruskal

Spanning Tree

- **Input:** Given a connected undirected graph G = (V, E)
- Output: A spanning tree of G is, i.e., a subset of edges that forms a tree and contains all the vertices in G.
- Applications
 - Building a network, connecting all hubs via minimum number of cables.
- Solutions
 - BFS, DFS.

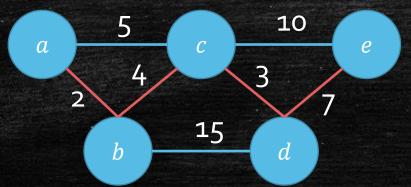
Minimum Spanning Tree

- **Input:** Given a connected undirected graph G = (V, E), and a weight function w(e) for each $e \in E$.
- Output: A spanning tree of G is, i.e., a subset of edges, with minimized total weight.
- Applications
 - Building a network, connecting all hubs via minimum number of cables.



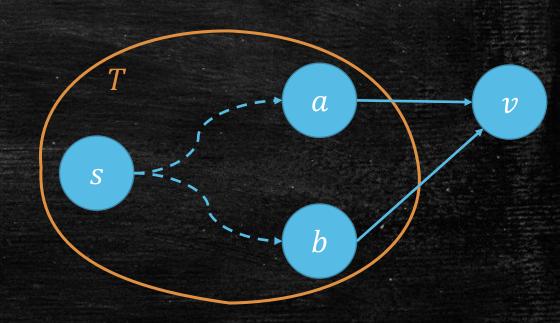
Minimum Spanning Tree

- Input: Given a connected undirected graph G = (V, E), and a weight function w(e) for each $e \in E$.
- Output: A spanning tree of G is, i.e., a subset of edges, with minimized total weight.
- Applications
 - Building a network, connecting all hubs via minimum number of cables.



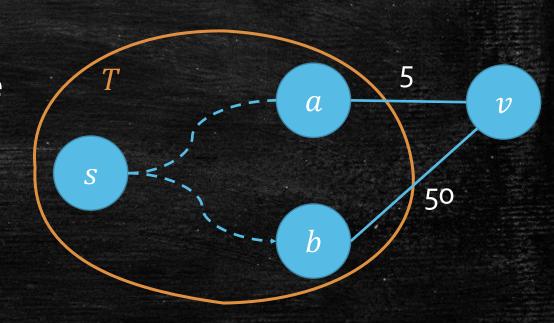
Dijkstra's growing idea

- Given a small SPT,
- choose a proper vertex v to find a larger SPT.
- New Plan for MST:
- Given a small MST,
- choose a proper vertex v to find a larger MST.



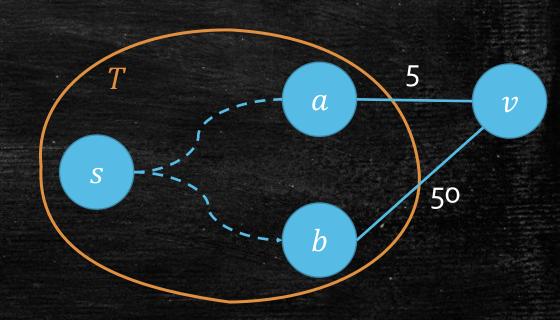
Prim's growing idea

- Given a small MST,
- choose a proper vertex v to find a larger MST.
- Which v is good?
- Dijkstra: v with smallest T-distance to s.
- Now: v with smallest cost!



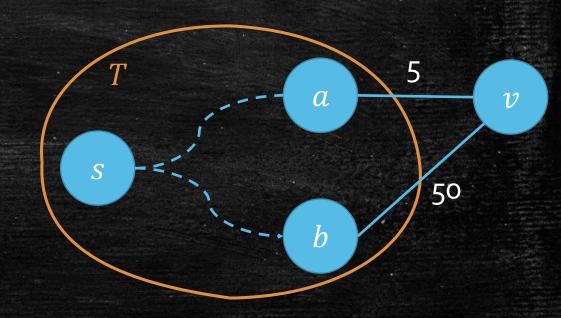
Prim's growing idea

- Given a small MST,
- choose a proper vertex v to find a larger MST.
- Grow v with smallest cost!
- Is it correct?
- Challenge:
 - How to define small MST

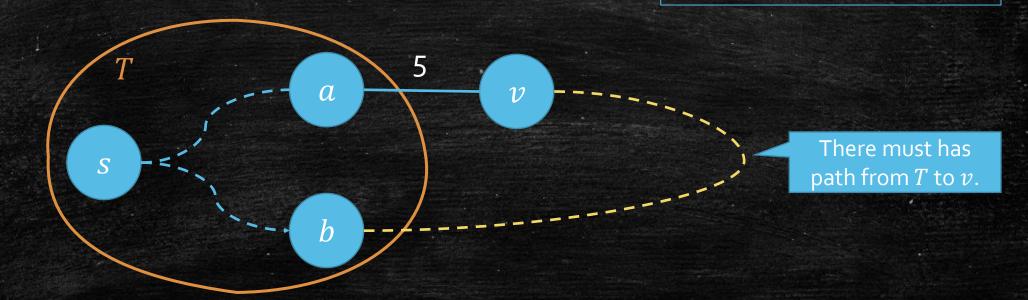


How to define small MST?

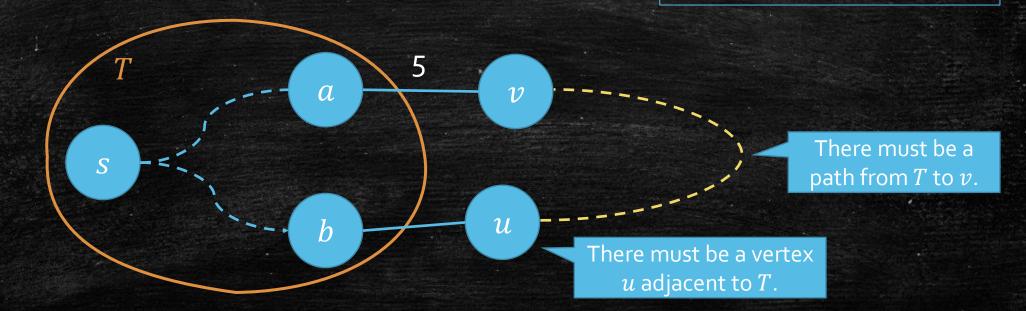
- T = (V', E') is a small MST if it is an MST for V'.
- Problem
 - are those edges in T still ok?
- A better choice:
- T is a P-MST (Partial MST) if it is a part of a complete MST for G.



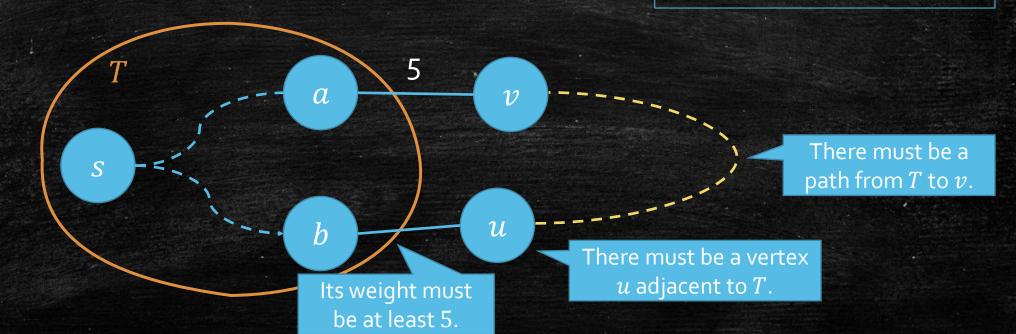
- **Given:** a small P-MST T.
- Want: a larger P-MST.
- Can we explore v (smallest cost) into T?



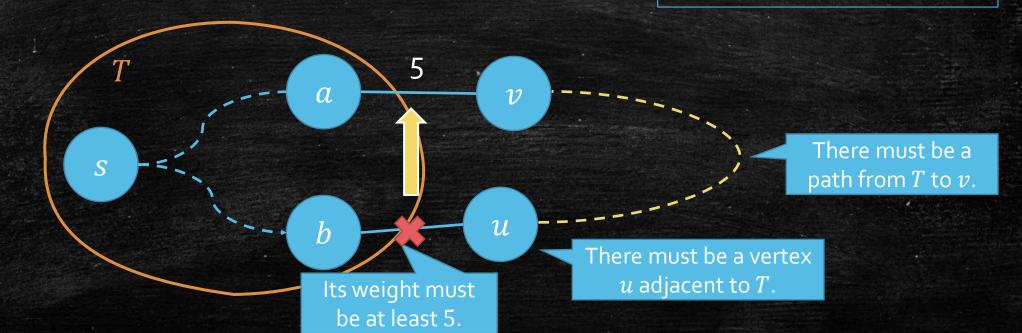
- **Given:** a small P-MST T.
- Want: a larger P-MST.
- Can we explore v (smallest cost) into T?



- **Given:** a small P-MST T.
- Want: a larger P-MST.
- Can we explore v (smallest cost) into T?



- **Given:** a small P-MST T.
- Want: a larger P-MST.
- Can we explore v (smallest cost) into T?

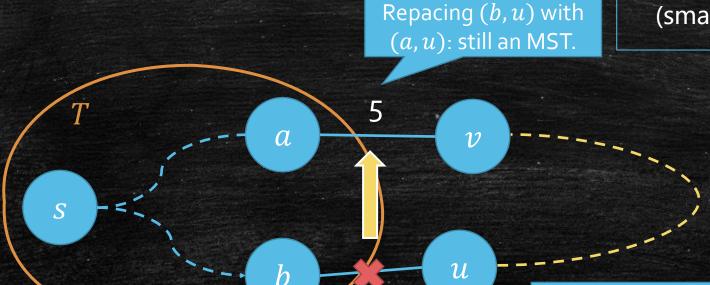


• Let's say T^* is the complete MST that contains T, and suppose $(a, v) \notin T^*$.

Given: a small P-MST T.

Want: a larger P-MST.

Can we explore v
 (smallest cost) into T?



Its weight must

be at least 5.

There must be a path from T to v.

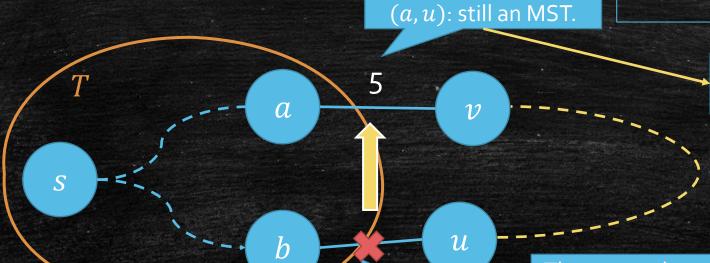
There must be a vertex u adjacent to T.

• Let's say T^* is the complete MST that contains T, and suppose $(a, v) \notin T^*$.

• **Given:** a small P-MST T.

Want: a larger P-MST.

Can we explore v
 (smallest cost) into T?



Its weight must

be at least 5.

Repacing (b, u) with

 $T \cup \{(a, v)\}$ must be a part of an MST.

There must be a path from T to v.

There must be a vertex u adjacent to T.

Prim Algorithm [Jarník '30, Prim '57, Dijkstra '59]

Prim(G = (V, E))

1. Initialize

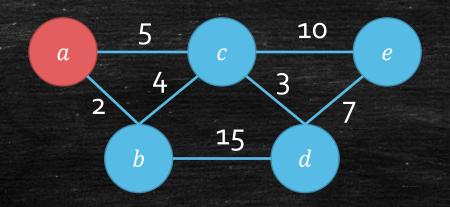
- $T \leftarrow {}$, S ← {s}; #s is an arbitrary vertex.
- cost[s] = 0, $cost[v] \leftarrow \infty$ for all v other than s.
- $-cost[v] \leftarrow w(s,v), pre[v] = s \text{ for all } (s,v) \in E.$

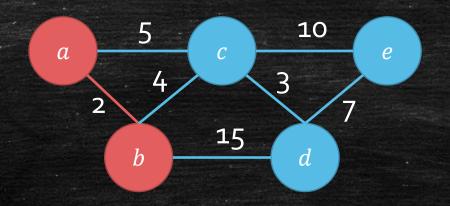
2. Explore

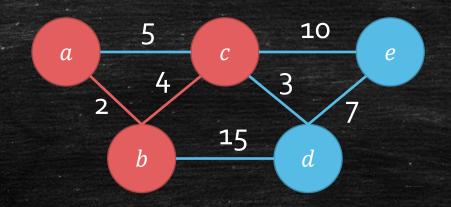
- Find $v \notin S$ with smallest cost[v].
- S ← S + {v}; T ← T + {(pre[v], v)}

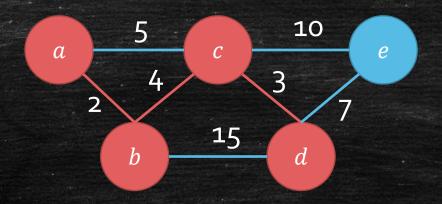
3. Update cost[u]

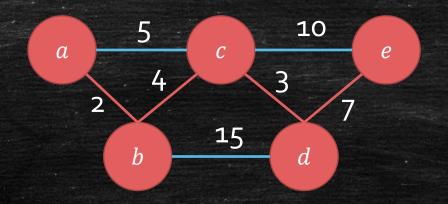
- $cost[u] = min\{cost[u], w(v, u)\}$ for all $(v, u) \in E$
- If cost[u] is updated, then pre[u] = v.











Running Time

- I believe you know how to analyze it:
- We can do it in $O(|E| + |V| \log |V|)$.

Kruskal Algorithm [Kruskal 1956]

Another Greedy!

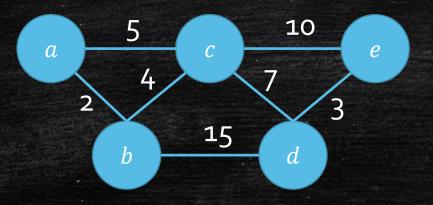
$$Kruskal(G = (V, E))$$

- Sort the edge set E to descending order.
- For each $e \in E$ in descending order
 - If e do not create a cycle, then choose it.

Kruskal Algorithm

Kruskal(G = (V, E))

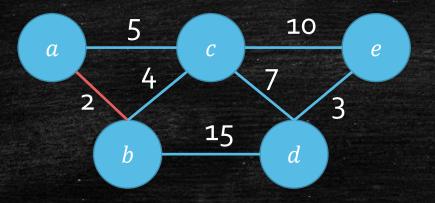
- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.



2 3 4 5 7 10 15

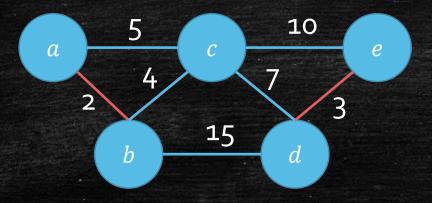
Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.



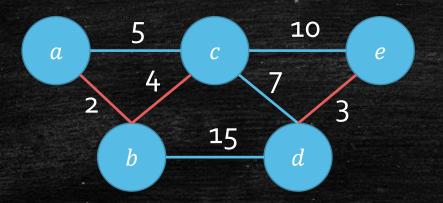
Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.



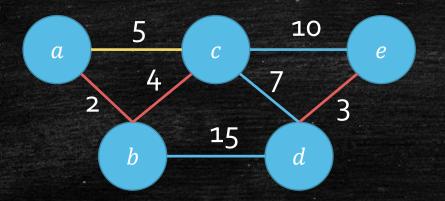
Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.



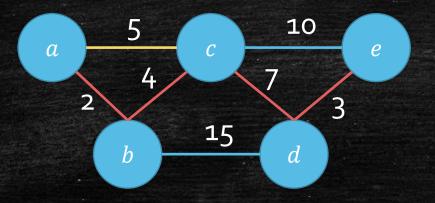
Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.



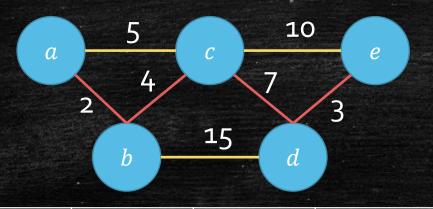
Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.



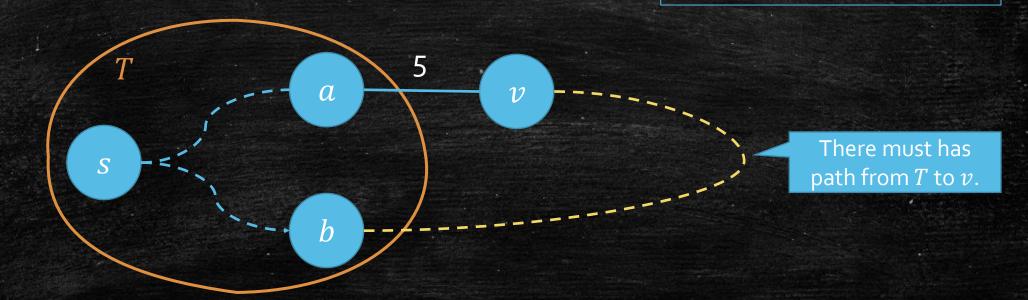
Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.



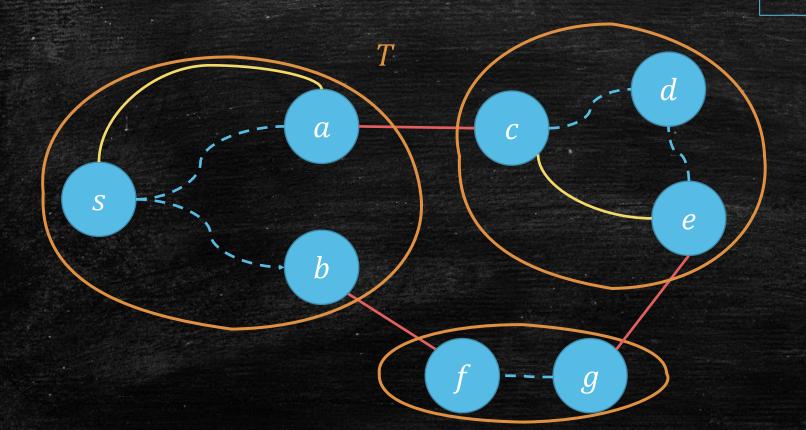
Correctness of Prim's Growing idea

- **Given:** a small P-MST T.
- Want: a larger P-MST.
- Can we explore v (smallest cost) into T?

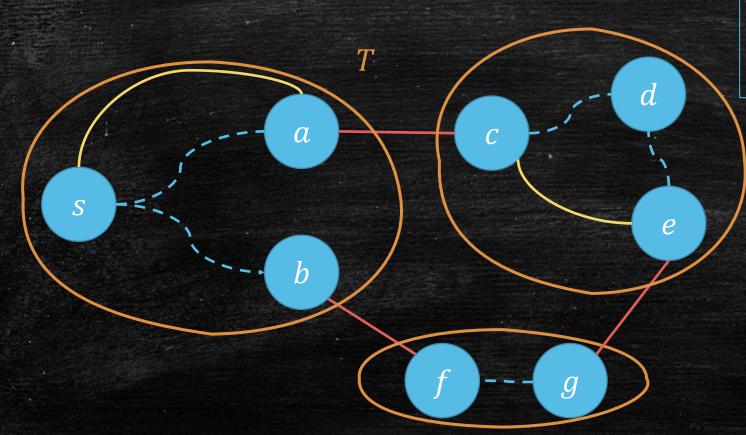


Correctness of Kruskal's Growing idea

- Given: a small P-MST T.
- Want: a larger P-MST.

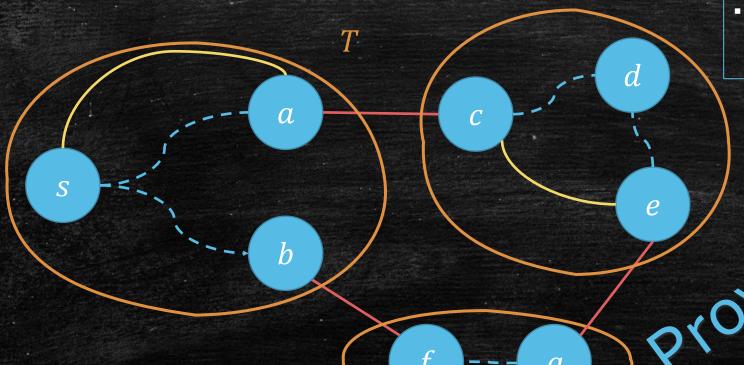


Correctness of Kruskal's Growing idea



- **Given:** a small P-MST T.
- Want: a larger P-MST.
- Add the smallest red edge get a larger P-MST.

Correctness of Kruskal's Growing idea



- Given: a small P-MST T.
- Want: a larger P-MST.
- Add the smallest red edge get a larger P-MST

Running Time

Kruskal(G = (V, E))

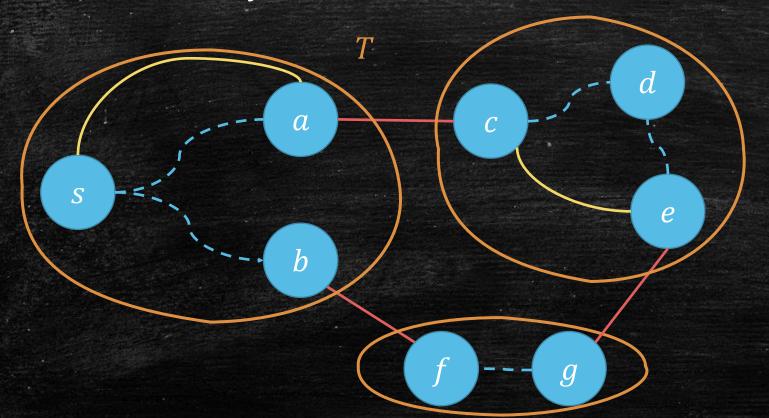
- Sort the edge set E to ascending order.
- For each $e \in E$ in ascending order
 - If e do not create a cycle, then choose it.
- $O(|E|\log|E|)$ for sorting.
- |E| round: check cycle!

Recall DFS

- When an edge is a back edge (to marked vertices),
- It forms a cycle.

During Kruskal

- When an edge connect the same group vertices,
- It forms a cycle.



Kruskal (refine)

Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $(u, v) \in E$ in ascending order
 - If group(u)! = group(v)
 - Choose (*u*, *v*).
 - union(group(u), group(v))

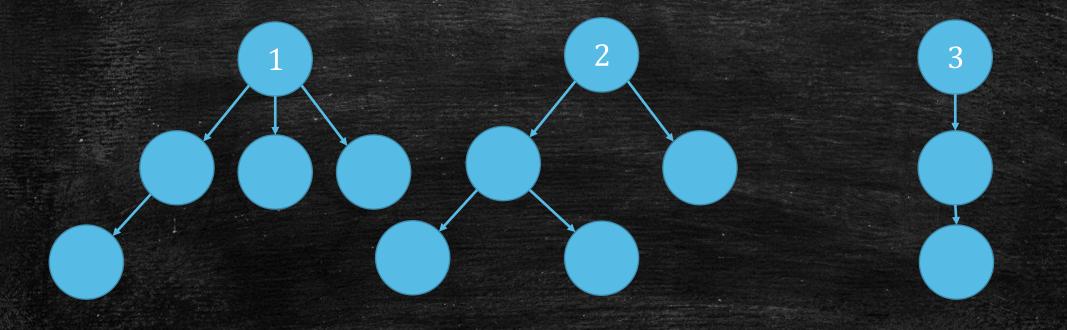
Running Time: Kruskal (refine)

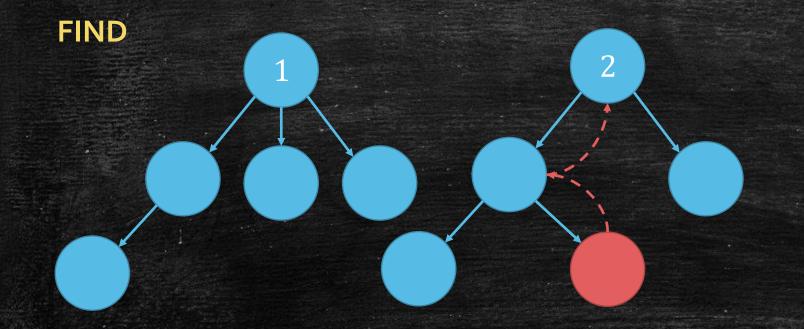
Kruskal(G = (V, E))

- Sort the edge set E to ascending order.
- For each $(u, v) \in E$ in ascending order
 - If group(u)! = group(v)
 - Choose (*u*, *v*).
 - union(group(u), group(v))
- $O(|E|\log|E|)$ for sorting.
- 2|E| round: check group
- |V| round: union group

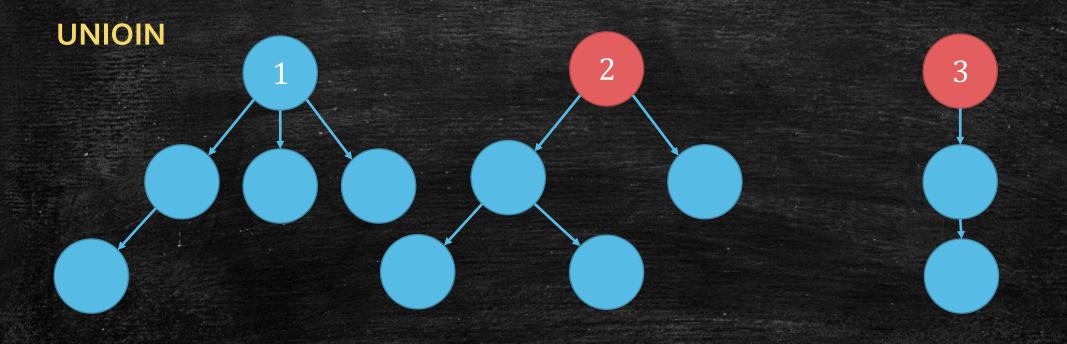
Union-Find Set

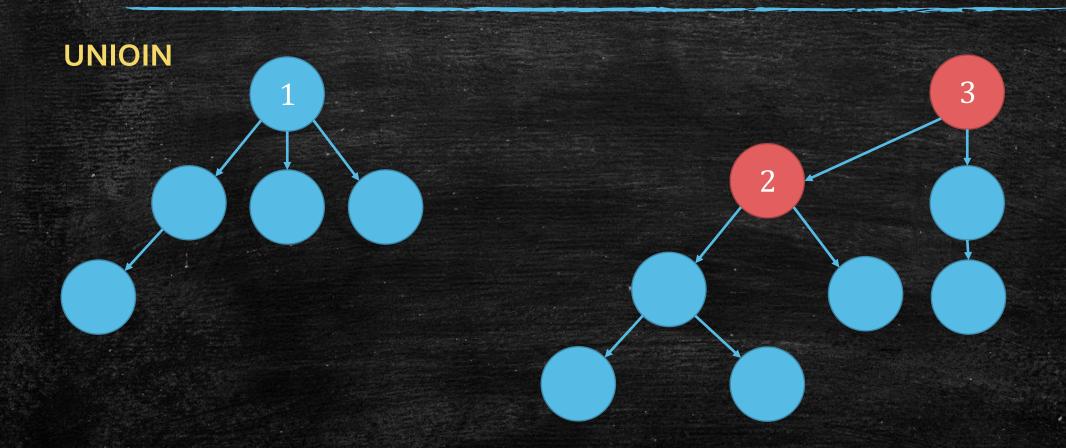
- Recall Union-Find Set
 - Find: $O(\log n)$
 - Union: 0(1)
- Kruskal
 - $O(|E|\log|E|)$ for sorting.
 - -2|E| round: check group
 - |V| round: union group
 - $O(|E|\log|E|) = O(|E|\log|V|)$







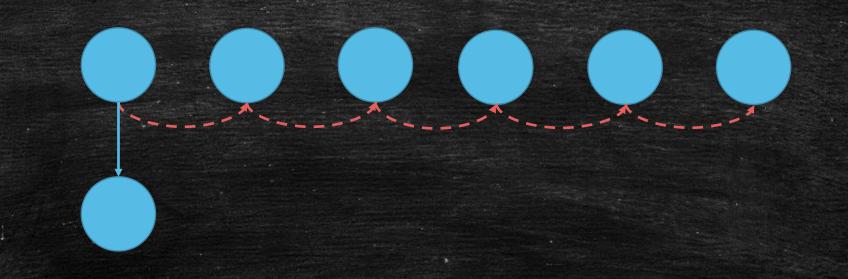


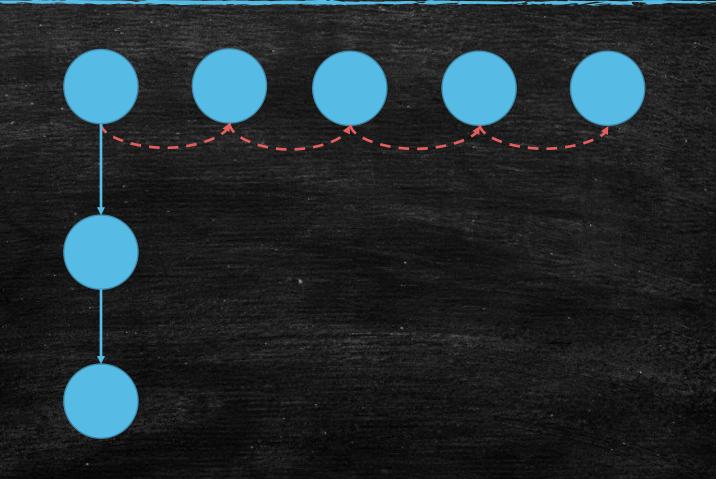


Time Complexity

- Find
 - O(max{Tree height})
- Union
 - 0(1)







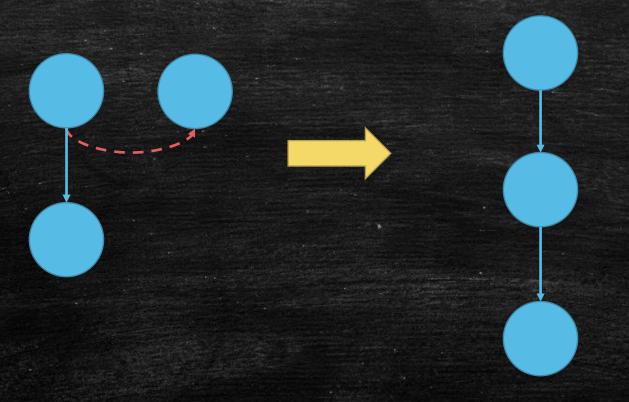
O(n) tree height

How to improve

- Find
 - O(max{Tree height})
 - O(n)!
- Union
 - 0(1)
- To Do
 - Reduce Tree Height

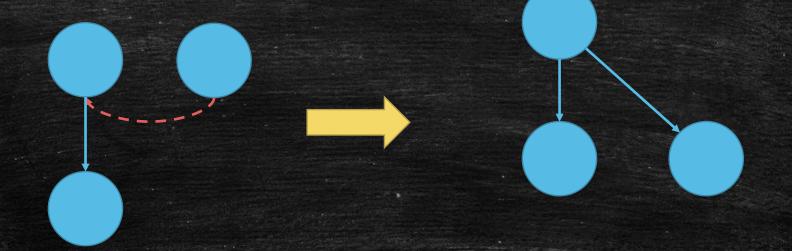
Intuition

BAD



Intuition

GOOD



We should merge to a same root!
We should merge short tree to high tree!

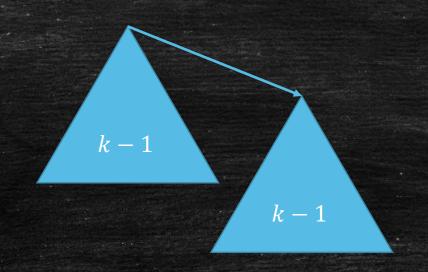
Implement

- Record Tree's height (rank).
- rank[v]: the rank of tree rooted at v.
- Union: u and v.
 - Rooted at u: if $rank[u] \ge rank[v]$
 - Rooted at v: if rank[u] < rank[v]
 - Update rank[u] + +: if rank[u] = rank[v]
- We make it hard to build a large rank tree!

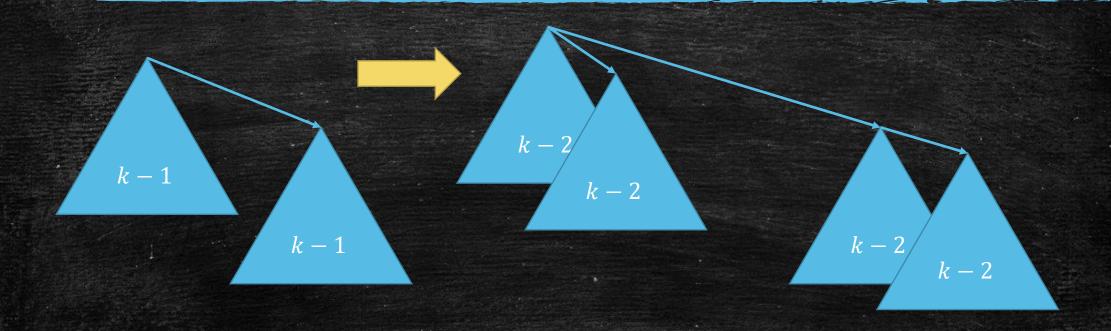
How to build a rank k tree?



How to build a rank k tree?



How to build a rank k tree?



We should at least use 2^k nodes!

Max tree height

- Build a rank k tree: We should at least use 2^k nodes!
- What is the max tree height (rank)?
- $O(\log n)$
- Find
 - O(max{Tree height})
 - $-O(\log n)!$
- Union (rank based)
 - 0(1)

Union-Find Set

- Recall Union-Find Set
 - Find: $O(\log n)$
 - Union: 0(1)
- Kruskal
 - $O(|E|\log|E|)$ for sorting.
 - -2|E| round: check group
 - |V| round: union group
 - $O(|E|\log|E|) = O(|E|\log|V|)$

Can we do better?

- Karger-Klein Tarjan (1995)
 - O(m) randomized algorithm.
- Chazelle (2000)
 - $O(m \cdot \alpha(n))$ deterministic algorithm.
 - $\alpha(n)$ is the inverse Ackermann function $\alpha(9876!)$ ≤ 5.
 - Ackermann function: $A(4,4) \approx 2^{2^{2^{16}}}$.
- Pettie-Ramachandran (2002)
 - O(optimal #comparison to determine solution)
 - We know #comparison = $\Omega(n) = O(m \cdot \alpha(n))$

Can we do better for Union-Find Set?

Have you heard Path Compression?

Path Compression

FIND

We put every red vertices to the first level.

Path Compression

FIND

We put every red vertices to the first level.

Path Compression

FIND

We put every red vertices to the first level.

Good for next FIND!

You know what the next step!

Amortized Analysis

Time Complexity

- Find (Path Compression)
 - $O(\log^* n)$ [Hopcroft & Ullman 1973]
 - $\log^*(2^{2^{2^2}}) = \log^*(2^{65536}) = 5$
 - $O(\alpha(n))$ [Tarjan 1975]
 - $\alpha(n)$ is the inverse Ackermann function $\alpha(9876!) = 5$.
- Union (rank based)
 - -0(1)

Rank Based Union + Find with Path Compression

- It is still an amortized analysis
- We prove:
 - m find operation, totally cost $O(m \log^* n)$.

Analysis

FIND

Cost=number of red edges.

Key Idea: Charge Cost to Vertices

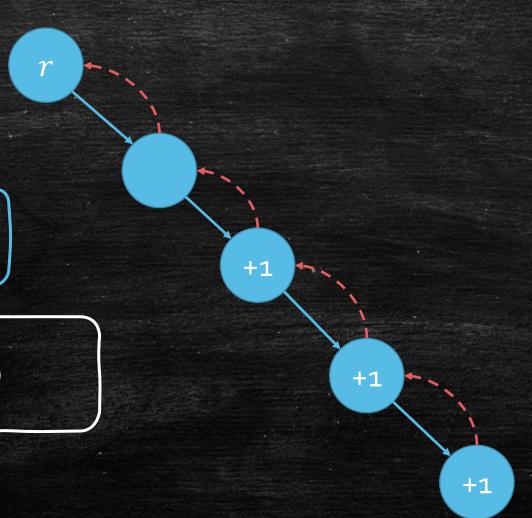
FIND

Cost=number of red edges.

Red edge charge cost to child vertex.

Total Cost of *m* FIND

- O(m)(to root)
- Total Charging



How much each vertex will be charge?

Group Vertices

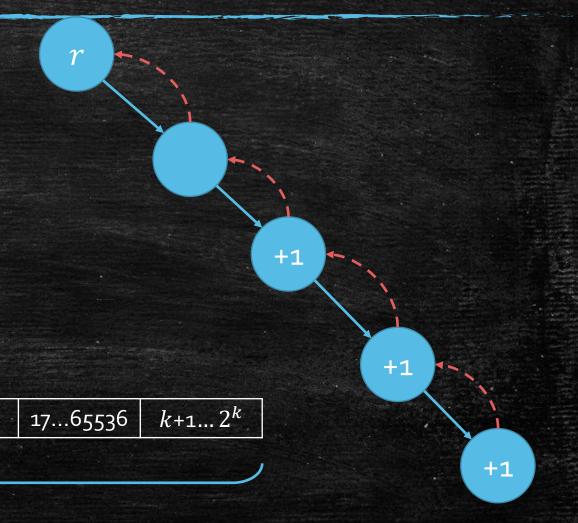
- rank(u) (slightly changed)
 - The rank of u when it become a child. (largest)
- Group vertices by rank



Different Type Charging

- Two kind of charging
 - Same Group Charging
 - Across Group Charing
- AGC for all vertices: $m \cdot \log^* n$

2...2



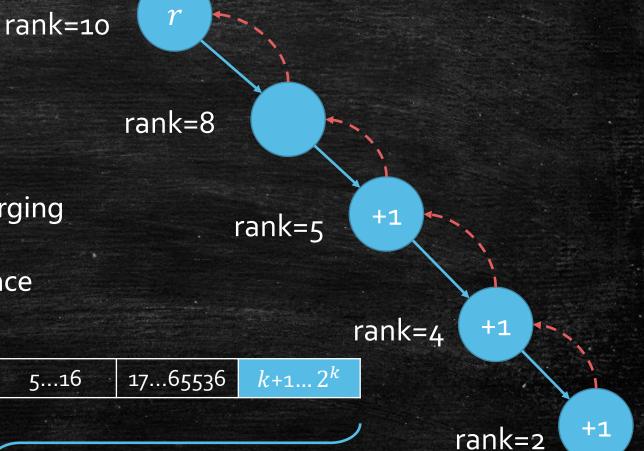
3...4

5...16

Bound SGC

- SGC for v
- One SGC
 - rank(parent(v)) > rank(v)
 - rank based union
 - rank(parent(v))++ after charging
 - path compression
 - For v, each rank charged once
 - Totally $2^k k \le 2^k$ charging

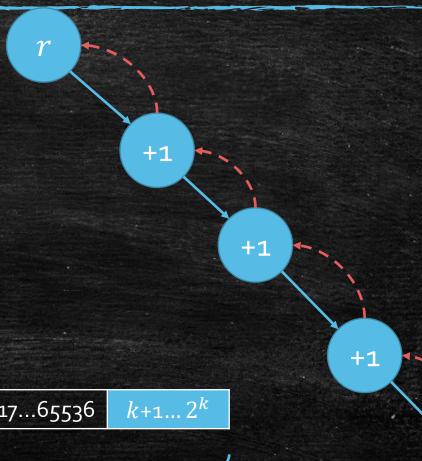




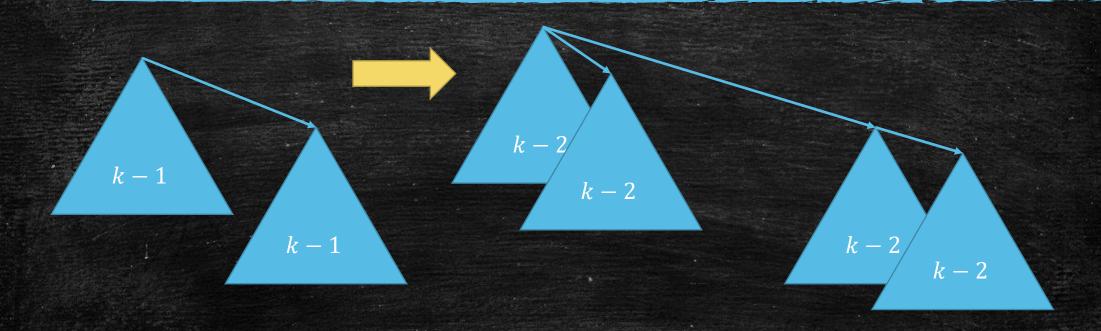
Bound Total cost

- Total Cost of m FIND
 - O(m)(to root)
 - Total AGC
 - $m \cdot \log^* n$
 - Total SGC
 - $\sum_{grops} #vertices \cdot 2^k$
 - #vertices $\leq n/2^k$
 - $n \log^* n$
 - Total : $m \log^* n$





Why #vertices $\leq \frac{n}{2^k}$?



We should at least use 2^k nodes!

Today's goal

- Learn what is Greedy!
- Learn to use Greedy to finish homework!
- Learn Prim and Kruskal!
- Again, how to use Data Structure to improve Algorithms.
- Review Union-Find Set!
- Learn another Amortized Analysis!