

The union-find data structure

COMS20010 (Algorithms II)

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A **union-find data structure** supports the following operations:

- **MakeUnionFind(X)**: Makes a new union-find data structure containing a 1-element set $\{x\}$ for each element $x \in X$. Takes $O(|X|)$ time.
- **Union(x, y)**: Merge the set containing x with the set containing y into a single set in the data structure. Takes $O(\log |X|)$ time.
- **FindSet(x)**: Returns a unique identifier for the set containing x . Takes $O(\log |X|)$ time.

Set identifiers can be anything as long as they're unique.

If we implement the sets as linked lists, then **FindSet** is too slow. If we implement them as arrays, then **Union** is too slow.

We'll take the pointer structure of a linked list to make **Union** fast, but arrange it differently to make **FindSet** fast as well.

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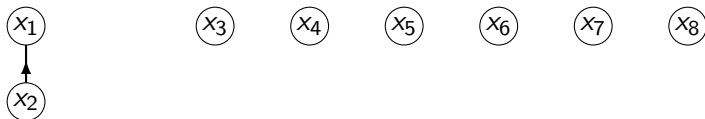


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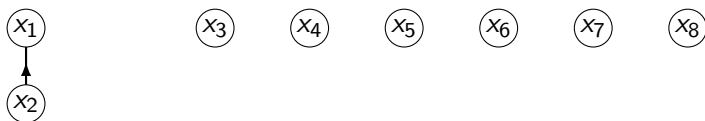


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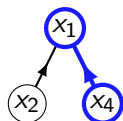
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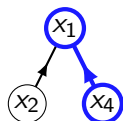
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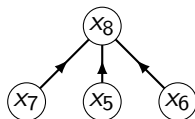
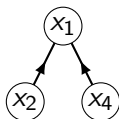


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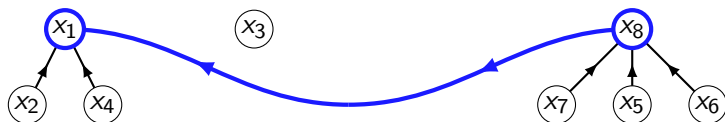


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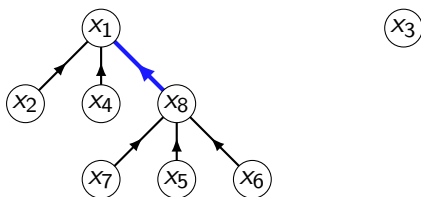


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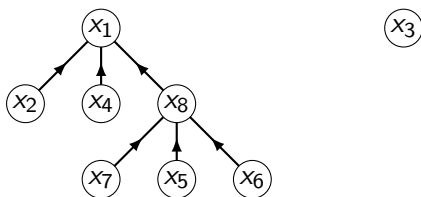


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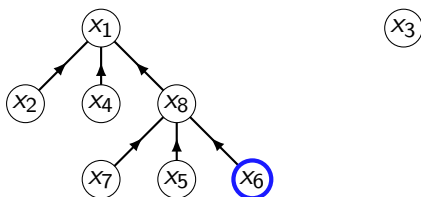


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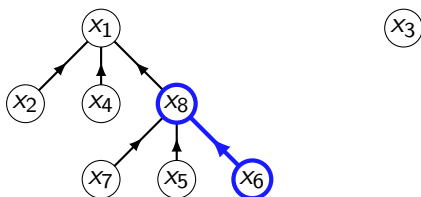


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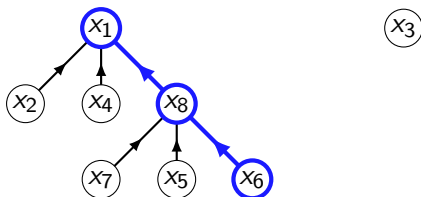


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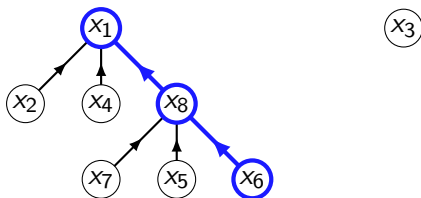
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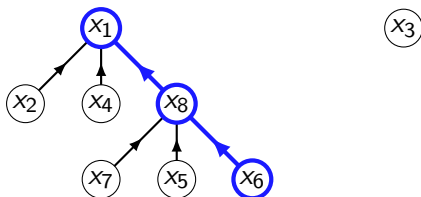
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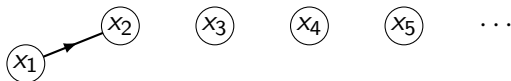
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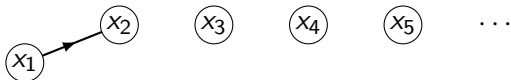
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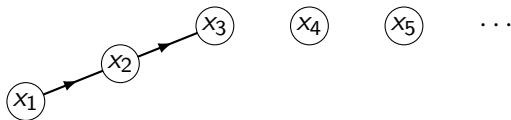
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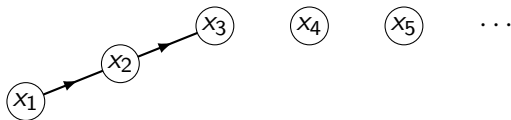
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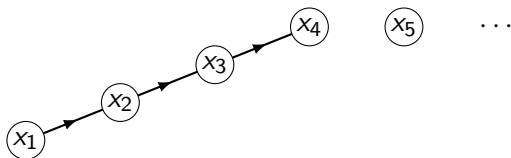
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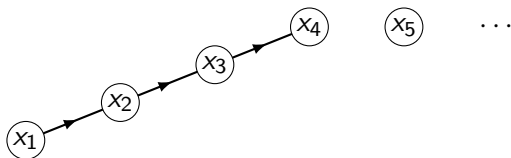
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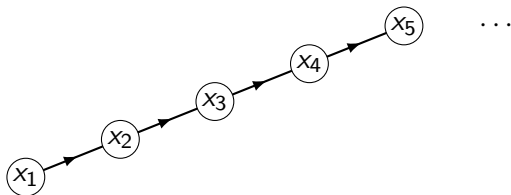
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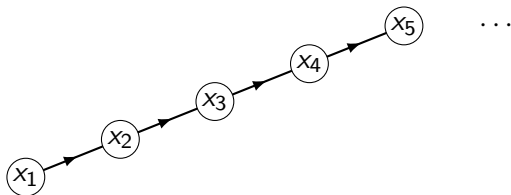
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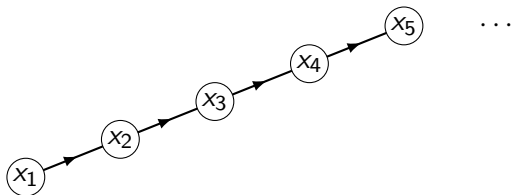
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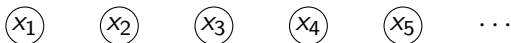
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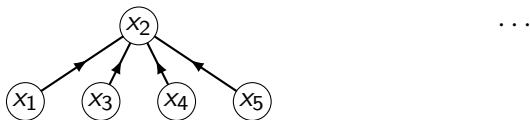
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This means any tree with depth greater than $\log |X|$ would contain more than $2^{\log |X|} = |X|$ vertices, which is impossible! So $d \leq \log |X|$.

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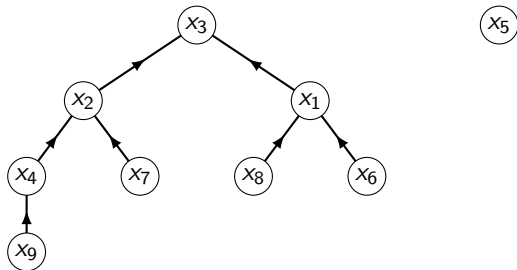
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In particular, we can use this to implement Kruskal's algorithm and Borůvka's algorithm in $O(|E| \log |E|)$ time!

A possible improvement: Path compression

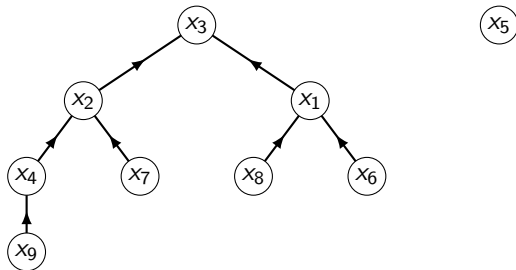
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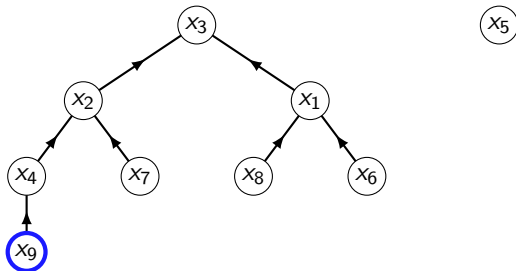
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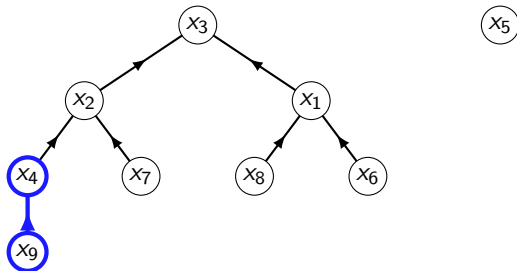
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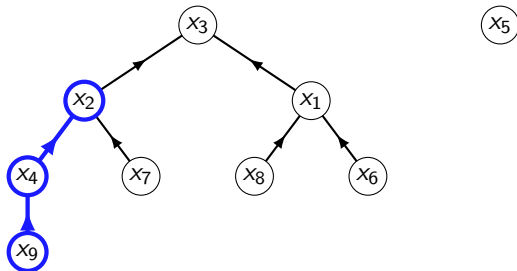
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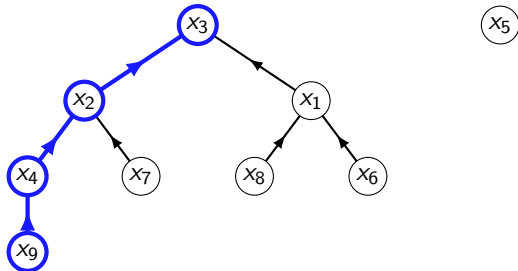
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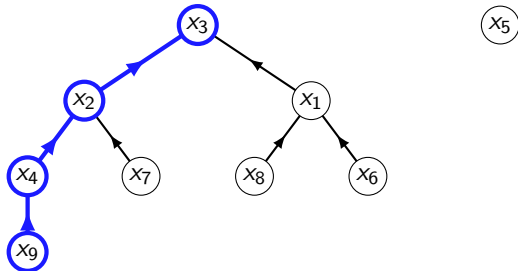


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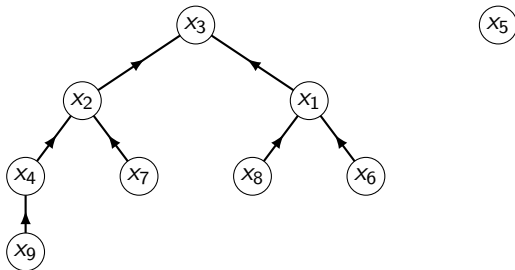
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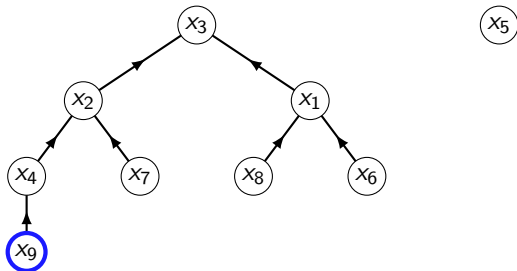
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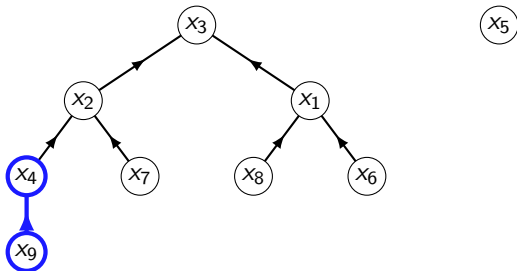
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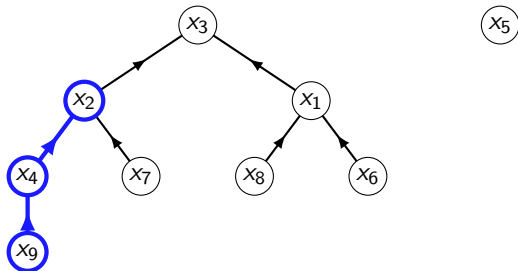
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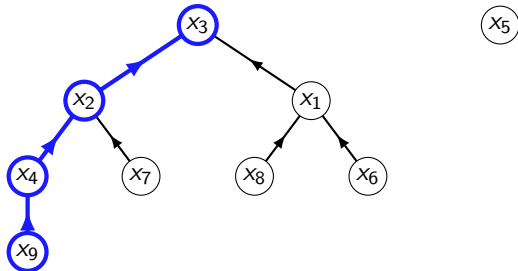
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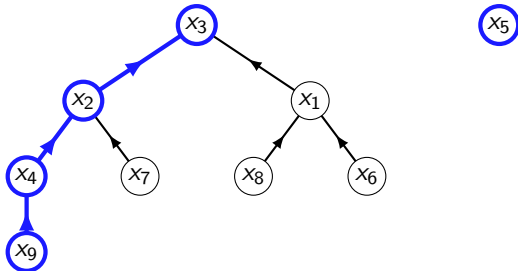
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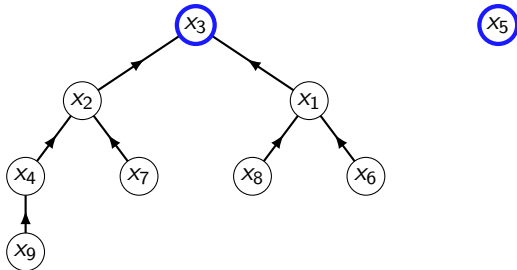
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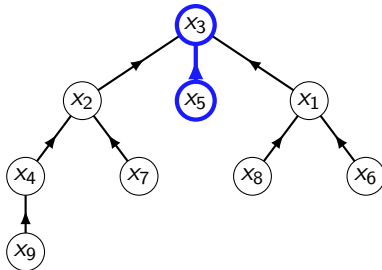
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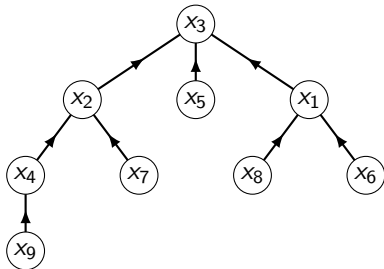
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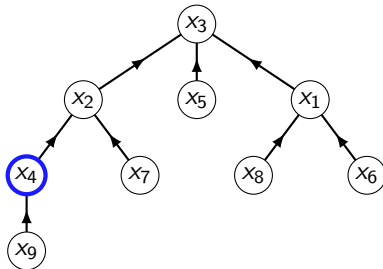
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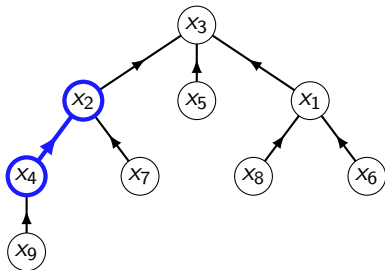
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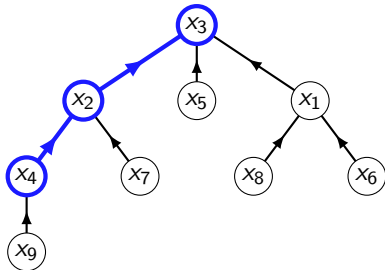
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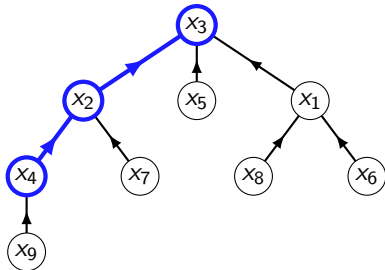


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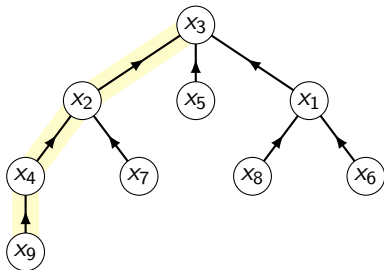
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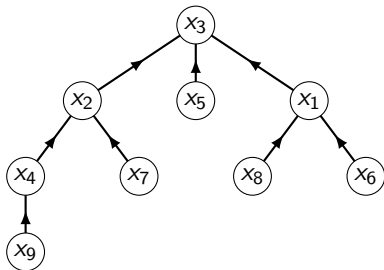
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We traverse these edges several times!



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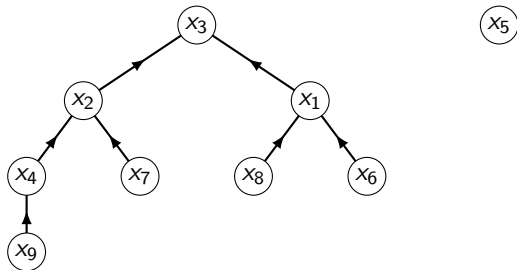


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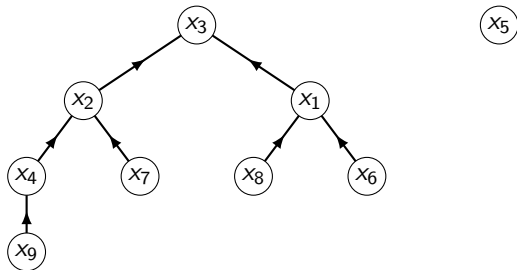
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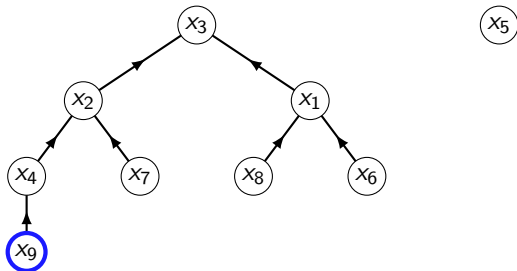
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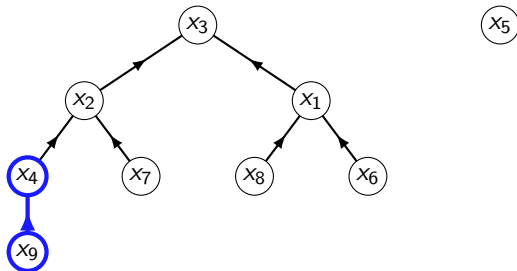
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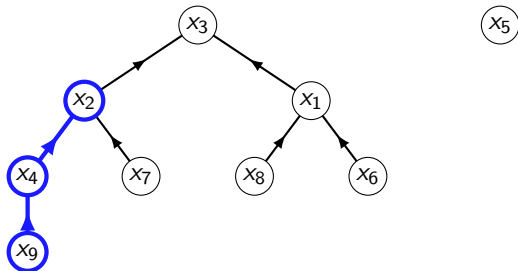
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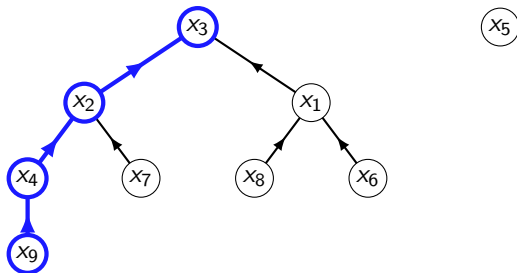
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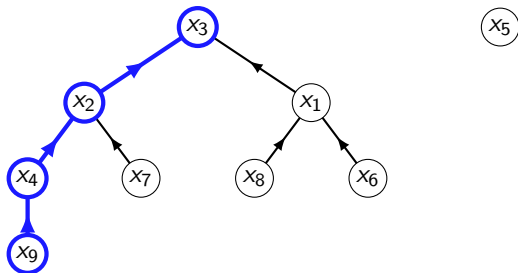
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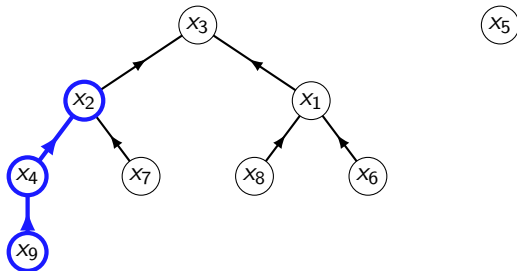
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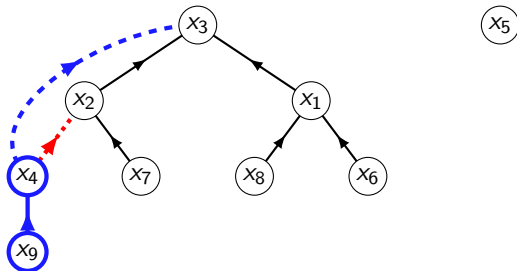
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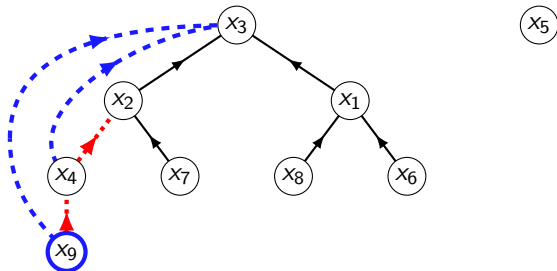
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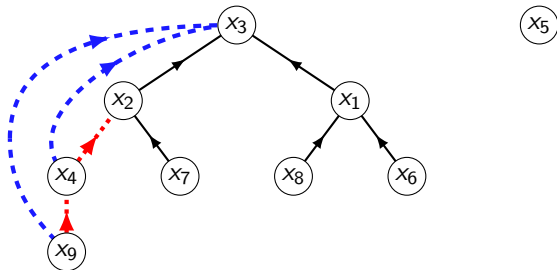
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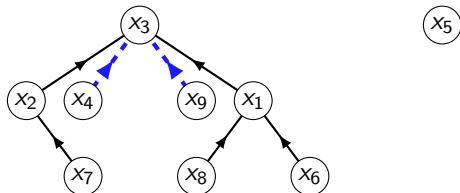
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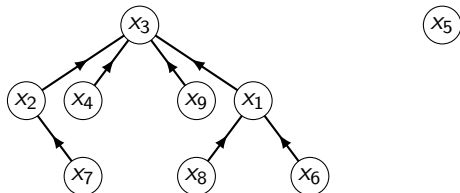
We could fix this by flattening our trees on each Union and FindSet operation, making every vertex we pass through a child of the root.

This technique is called **path compression**.

A possible improvement: Path compression

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`Union(x9, x5);`



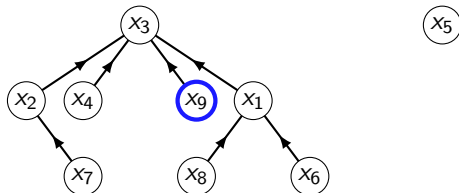
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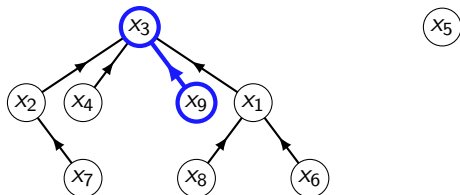
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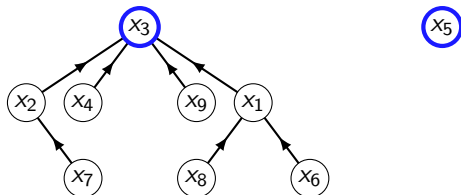
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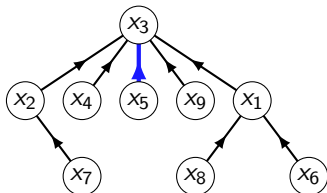
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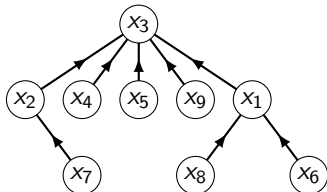
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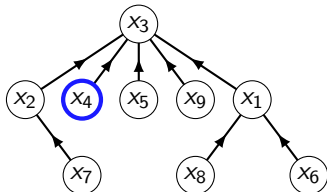
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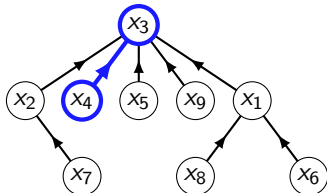
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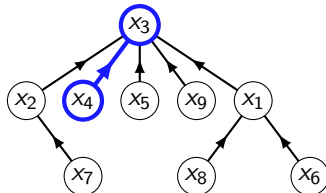
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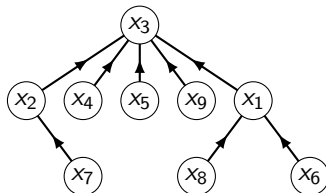
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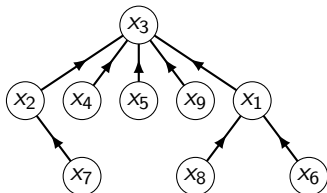
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This improves the running time of n operations from $O(n \log n)$ to $O(n\alpha(n))$, where $\alpha(n)$ is the **inverse Ackermann function**: $\alpha(n) = \min\{k: A(k, k) \geq n\}$.

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This improves the running time of n operations from $O(n \log n)$ to $O(n\alpha(n))$, where $\alpha(n)$ is the **inverse Ackermann function**: $\alpha(n) = \min\{k: A(k, k) \geq n\}$. In practice, we **always** have $\alpha(n) \leq 4$.