The union-find data structure COMS20010 2020, Video 6-2

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Last time...

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.

We will implement the data structure not as a set of linked lists, but as a **forest** in which the elements are vertices and the sets are **components**.

MakeUnionFind $(x_1, x_2, x_3, x_4, x_5, x_6, x_7, x_8)$;

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 $FindSet(x_3);$

















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FindSet (x_3) ; Returns x_3 .

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Union
$$(x_1, x_2)$$
;









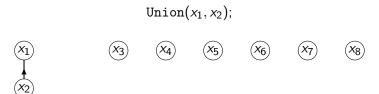




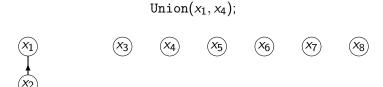




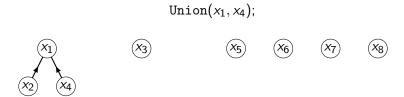
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 (x_3)



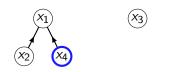






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(X₃)

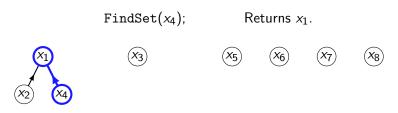


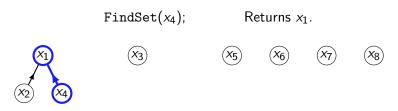






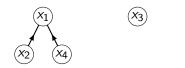
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- MakeUnionFind(X) makes an isolated vertex for each element of X.
- FindSet(x) returns the root of x's tree as its identifier.

Union
$$(x_5, x_8)$$
; Union (x_8, x_7) ; Union (x_6, x_8) ;











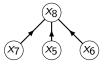
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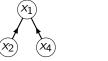


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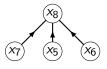


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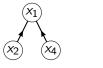




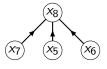


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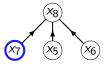


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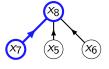
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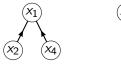


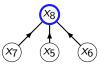
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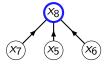
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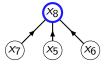
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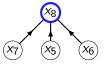
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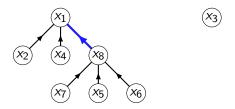
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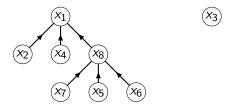
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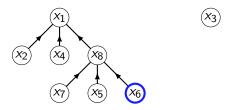
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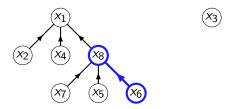
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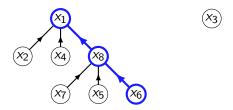
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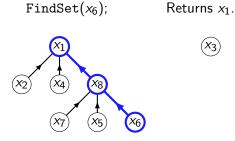


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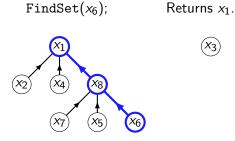


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Forests can degenerate into linked lists

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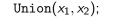






$$(X_4)$$

$$(x_5)$$









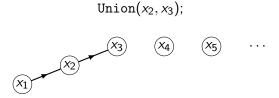


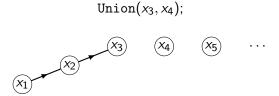






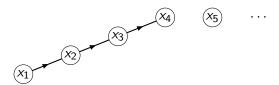






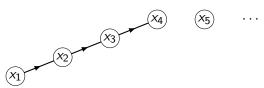
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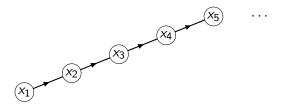
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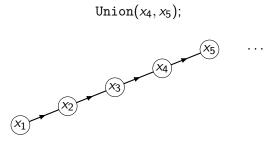


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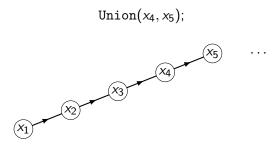
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Recall Union (x_i, x_j) puts the root of x_i under the root of x_j , or vice versa.

If we make bad choices of which root goes under which, like the above, we may have $d \in \Theta(|X|)$. How can we prevent this?

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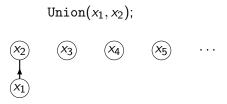




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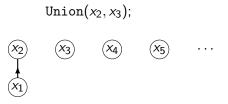
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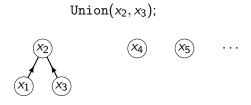
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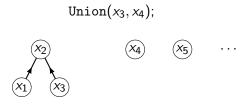
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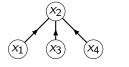


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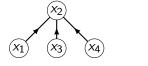
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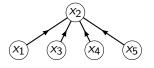
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Union (x_1, x_2) follows pointers from x_1 and x_2 up to the roots r_1 and r_2 of their trees. If x_1 's tree has lower depth than x_2 's tree, then it adds r_1 as a child of r_2 ; if not, it adds r_2 as a child of r_1 .

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Writing d_1 for the depth of x_1 's tree, and d_2 for the depth of x_2 's tree, if $d_1 < d_2$ then the depth of the new tree will be $\max\{d_2, d_1 + 1\} = d_2$.

Likewise, if $d_2 < d_1$ then the new depth will be $\max\{d_1, d_2 + 1\} = d_1$.

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Proof: By induction on *d*.

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The depth only increases if $d_1 = d_2$.

Lemma: If the data structure contains a tree of depth d, then it has at least 2^d vertices in total.

Proof: By induction on *d*.

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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 $d \le \log |X|$.

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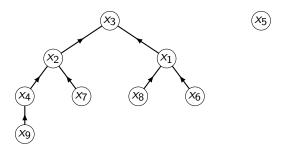
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In particular, we can use this to implement Kruskal's algorithm and Borůvka's algorithm in $O(|E|\log|E|)$ time!

A possible improvement

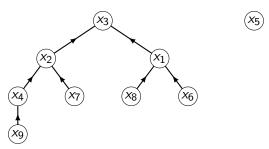
Right now, we are duplicating some work with root-finding.



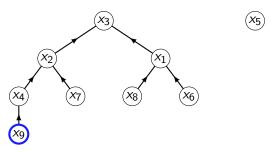
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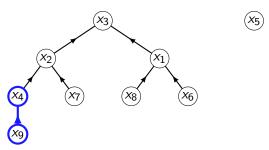
 $FindSet(x_9);$



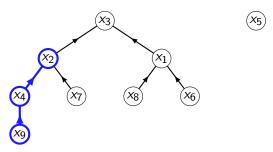
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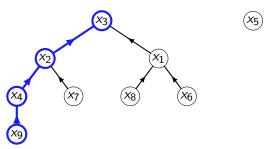
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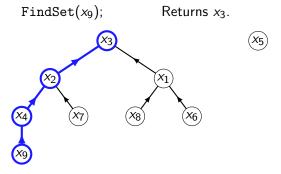


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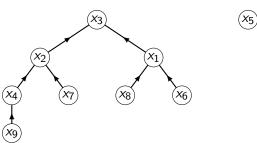
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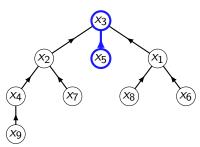
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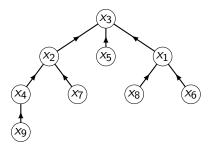
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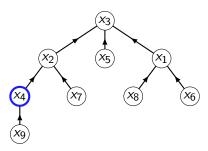
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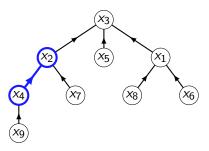


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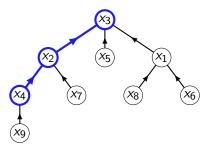


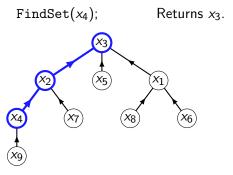
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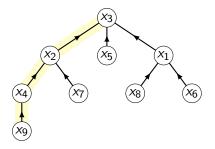
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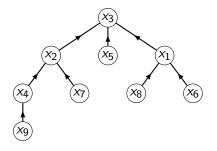


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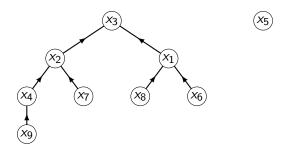
We traverse these edges several times!



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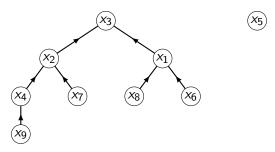


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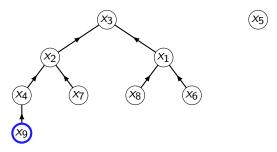
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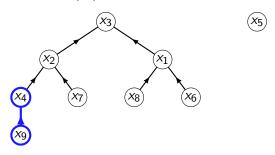
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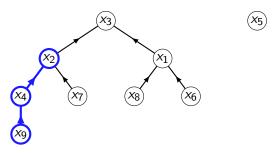
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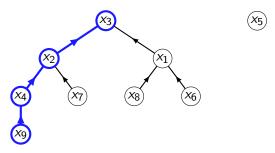
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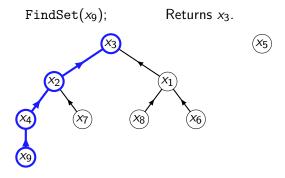


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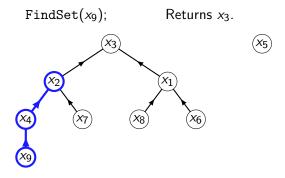
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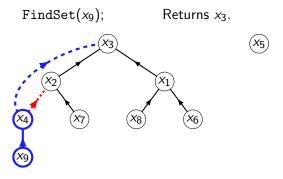
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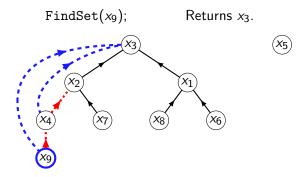
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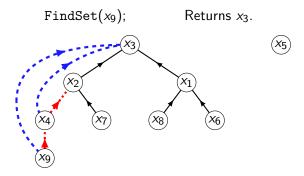
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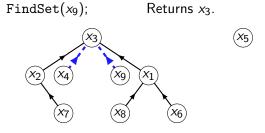
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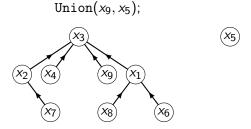
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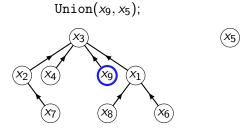
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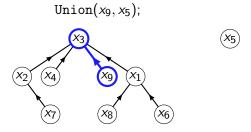
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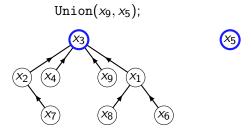
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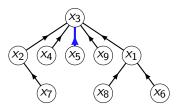


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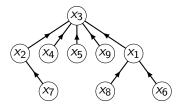
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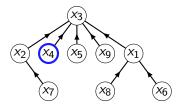
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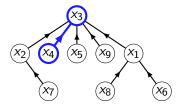
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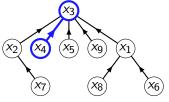
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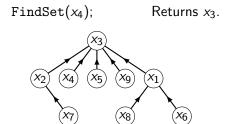


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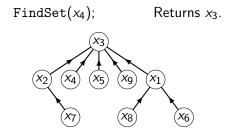
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This improves the running time of n operations from $O(n \log n)$ to $O(n\alpha(n))$, where $\alpha(n)$ is the **inverse Ackermann function**: $\alpha(n) = \min\{k \colon A(k,k) \ge n\}.$

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John Lapinskas Video 6-2 7/7