Efficient top-down updates in AVL trees

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- Balanced binary search trees
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- 4 Conclusion

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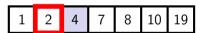


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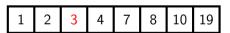
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Add 3!



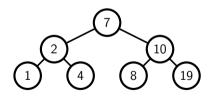
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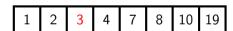
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Use a low-height binary search tree instead.



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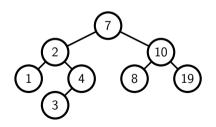
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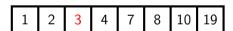
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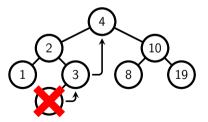
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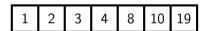
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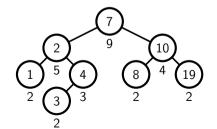
Remove 7!



• Weight-bounded trees:
$$\mathbf{r}(x) = \#\mathcal{T}(x) + 1$$

 $\mathbf{r}(x_i) \leqslant \alpha \mathbf{r}(x)$

$$(\mathsf{r}(x) = \mathsf{r}(x_1) + \mathsf{r}(x_2) \text{ and } \mathsf{r}(\bot) = 1) \ (1/\sqrt{2} \leqslant \alpha < 9/11)$$

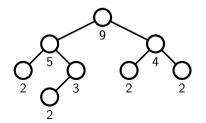




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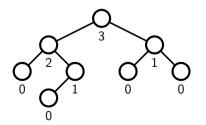
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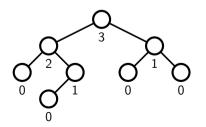


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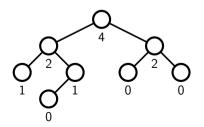


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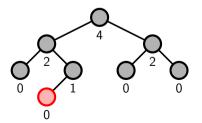


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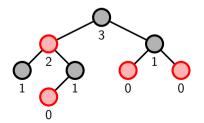


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Maintaining search trees of height $\mathcal{O}(\log(n))$ often requires some kind of rank and balance.

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Tree	Invented	Height	Am. writes/update	Top-down
Weight-balanced	1972	$2\log_2(n)$	Θ(1)	yes
AVL	1962	$1.44\log_2(n)$	$\Theta(\log(n))$	no
Weak AVL	2015	$2\log_2(n)$	Θ(1)	yes
Red-black	1978	$2\log_2(n)$	Θ(1)	yes

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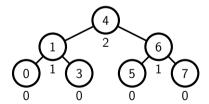
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In weak AVL trees, the first q queries trigger at most 13q rank updates and/or pointer rewrites: Their amortised write complexity is 13 operations per update.

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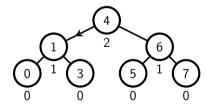
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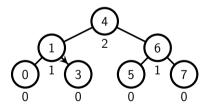
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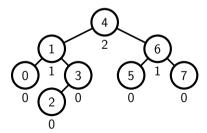
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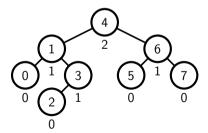
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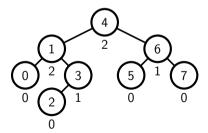
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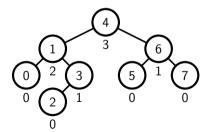
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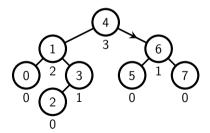
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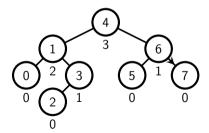
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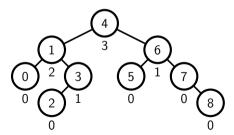
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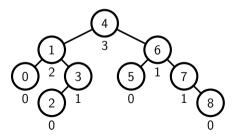
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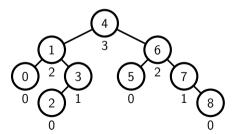
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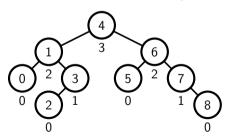


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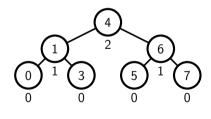
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Insert 2, then 8, bottom-up



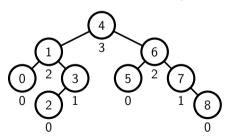
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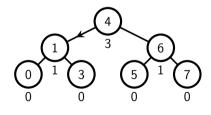
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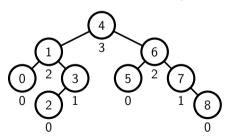
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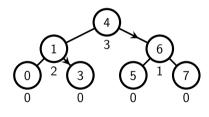
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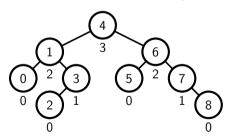
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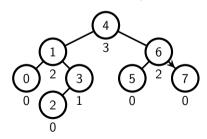
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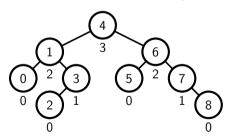
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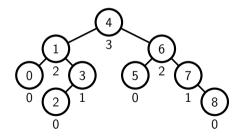
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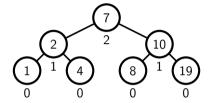


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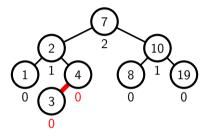


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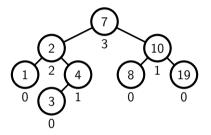


Goals: Avoiding zero-edges.



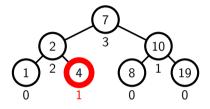


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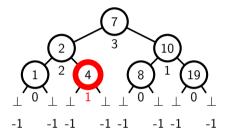


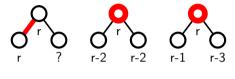


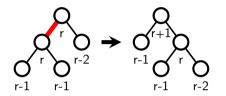
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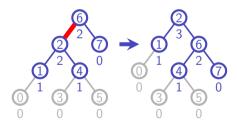


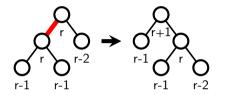


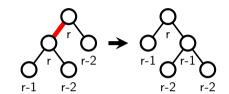


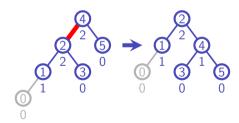


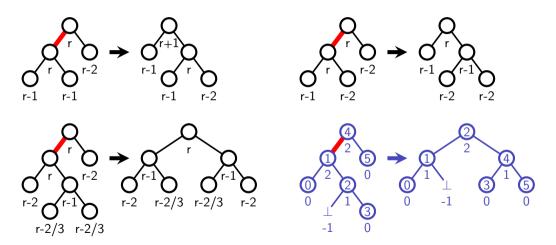


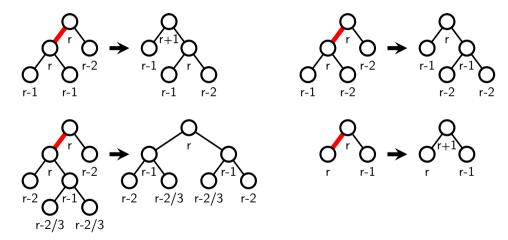


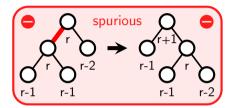


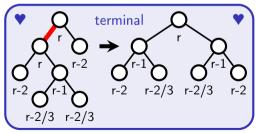


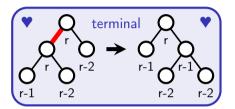


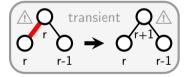


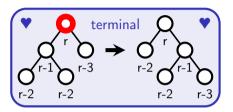


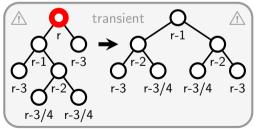


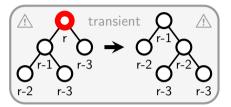


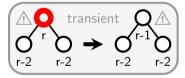


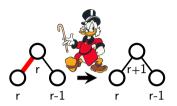


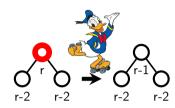


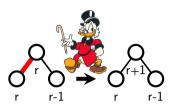


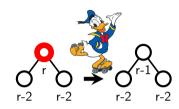


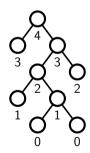


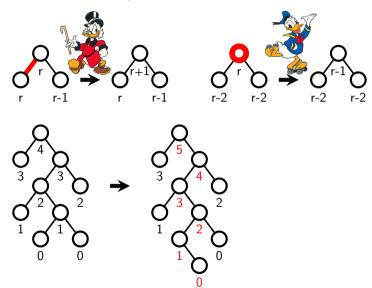


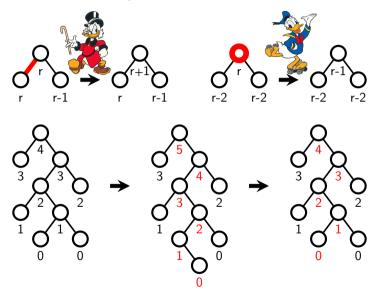








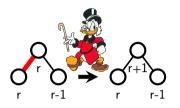


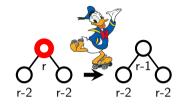


Updating AVL trees: The evil pair...and how to defeat it!



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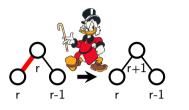


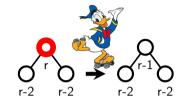






Updating AVL trees: The evil pair...and how to defeat it!









ACM Transactions on **Algorithms**

Article 30 (26 pages)

B. Haeupler S. Sen Rank-Balanced Trees

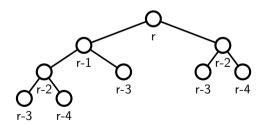
R. E. Tarjan

Stop propagating 0-edges and 4-nodes faster by demoting and promoting nodes.

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Hollow nodes can be demoted.

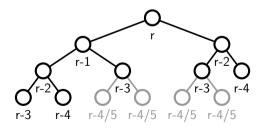
(descendants at rank $\geqslant r-2$ have a 2-child)



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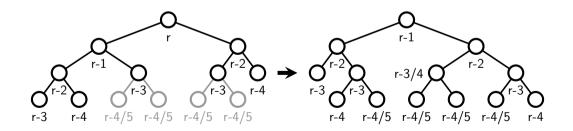
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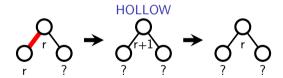
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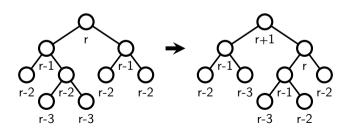
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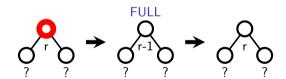
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Using our fast stopping procedure,

Transient deletion operations create no full nodes and destroy one 2-child. insertion operations destroy one full node and create one 2-child.

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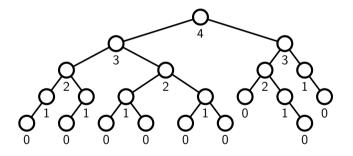


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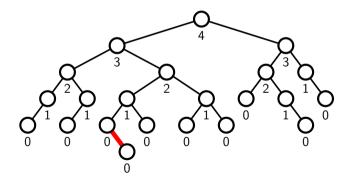
- Balanced binary search trees
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Safe nodes: When do we stop anomalies?

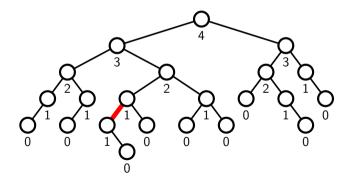
A node is insertion-safe if its stops propagating zero-edges.



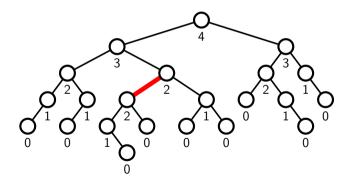




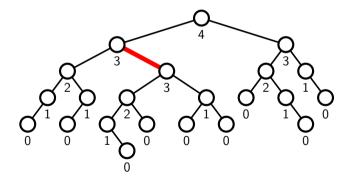




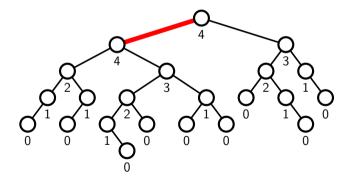




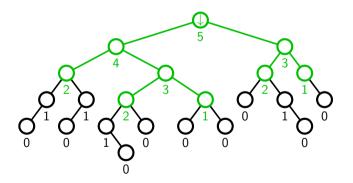




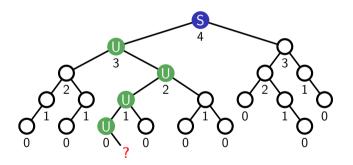










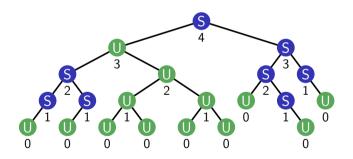




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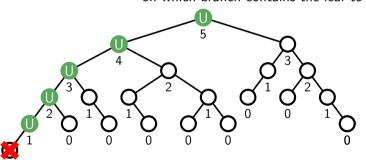
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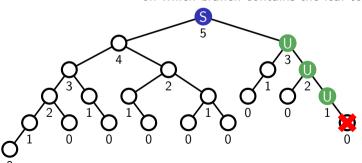
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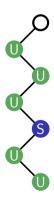
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Top-down insertion algorithm:

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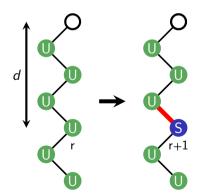
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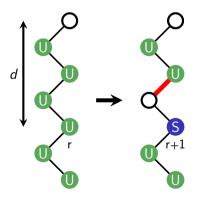
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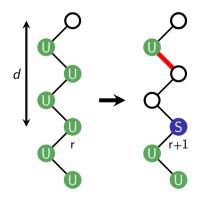
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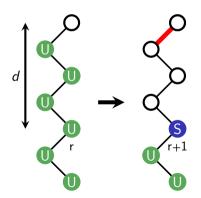
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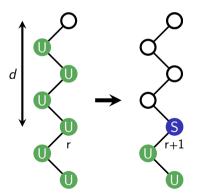
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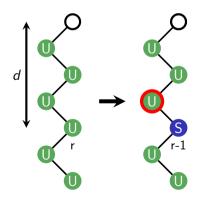
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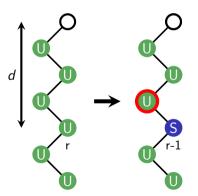
- 1 Look for a safe node on your deletion branch.
- If you succeed quickly, restart from that node. (write cost = 0)
- $\textbf{ § If you fail,} \qquad \qquad (\mathsf{write cost} = \mathcal{O}(d+1) \ \& \ \Delta \mathsf{Pot} \leqslant \mathcal{O}(1) d)$
 - demote an unsafe node at depth d & make it safe;
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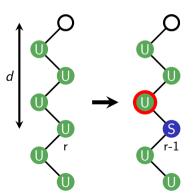
Challenge: Demote deletion-unsafe nodes & make them safe!



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Challenge: Demote deletion-unsafe nodes & make them safe! Answer: When facing 4 unsafe nodes in a row, you can do it!



Updating AVL trees efficiently top-down

Theorem

Starting from the empty AVL tree, q top-down queries trigger $\mathcal{O}(q)$ write operations.

Proof:

Tree potential decreases with each batch of d transient operations once d is large enough.

With q queries + B batches:

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Ongoing tasks:

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Main take-away:

Analyse algorithms and data structures that you love and adapt them!

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