

Lecture 5

Hashing (contd.), Binary Search Trees

Independent Uniform Hashing

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$$U = \{0, 1, \dots, n - 1\}$$

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What is the expected length of a chain now?

Probability Recap

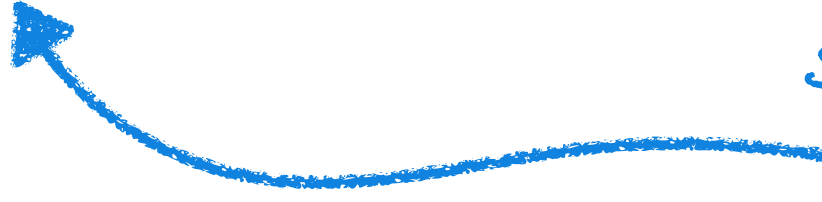
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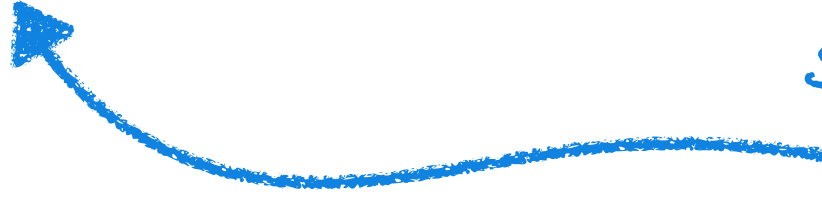


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X :	2	...	5	5	...	7	...	12

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Linearity of Expectation: $\mathbf{Ex}[X_1 + X_2 + \dots + X_n] = \mathbf{Ex}[X_1] + \mathbf{Ex}[X_2] + \dots + \mathbf{Ex}[X_n]$

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Search's Time Complexity: $O(1)$, when $n = m$.

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- Number of elements cannot exceed the size of hash table.

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Such that for all the keys k :

Probe sequence $\langle h(k, 0), h(k, 1), \dots, h(k, m - 1) \rangle$ is a permutation of $\langle 0, 1, \dots, m - 1 \rangle$

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9. **while** $i \neq m$
10. **error** “Hash table overflow”

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Flaw: Deletion like this can affect the search or deletion of previously inserted element.



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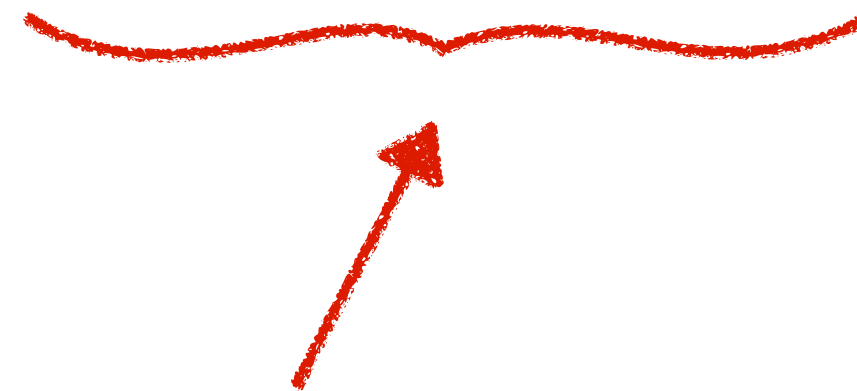
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To hit all the slots

Binary Search Trees

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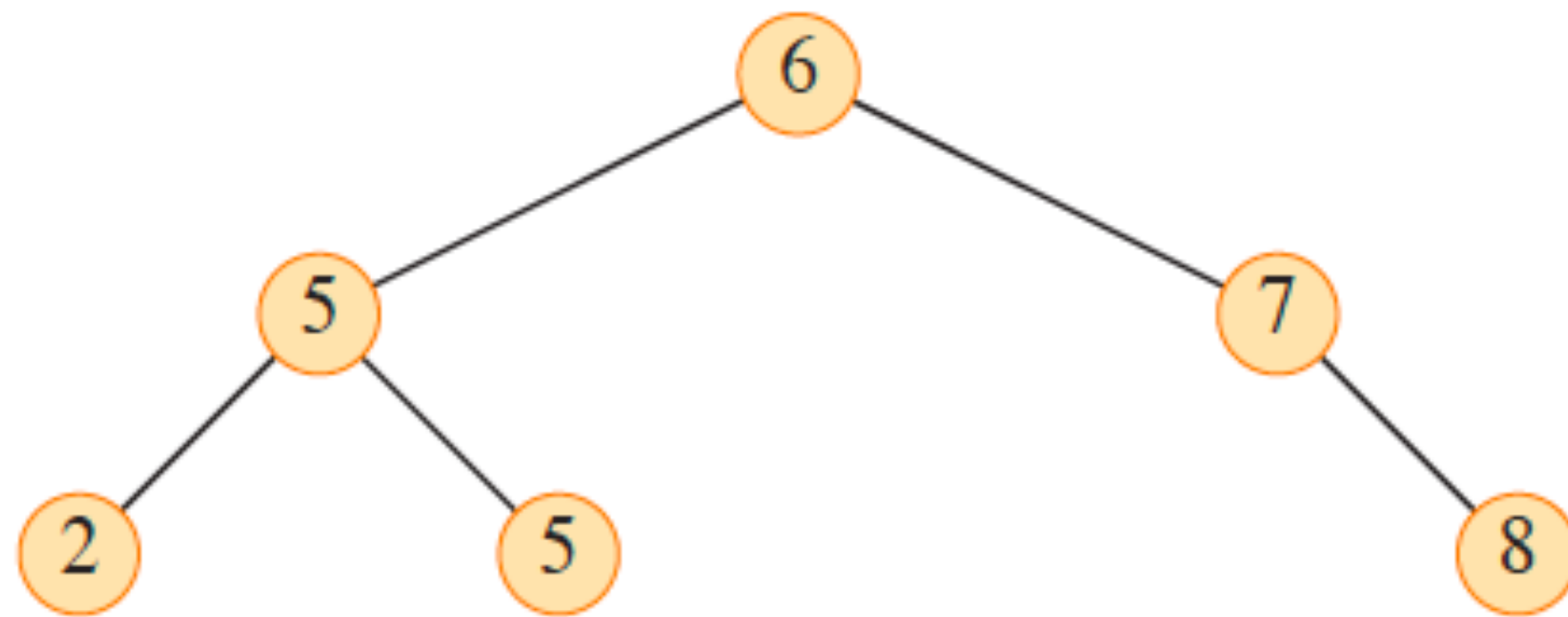
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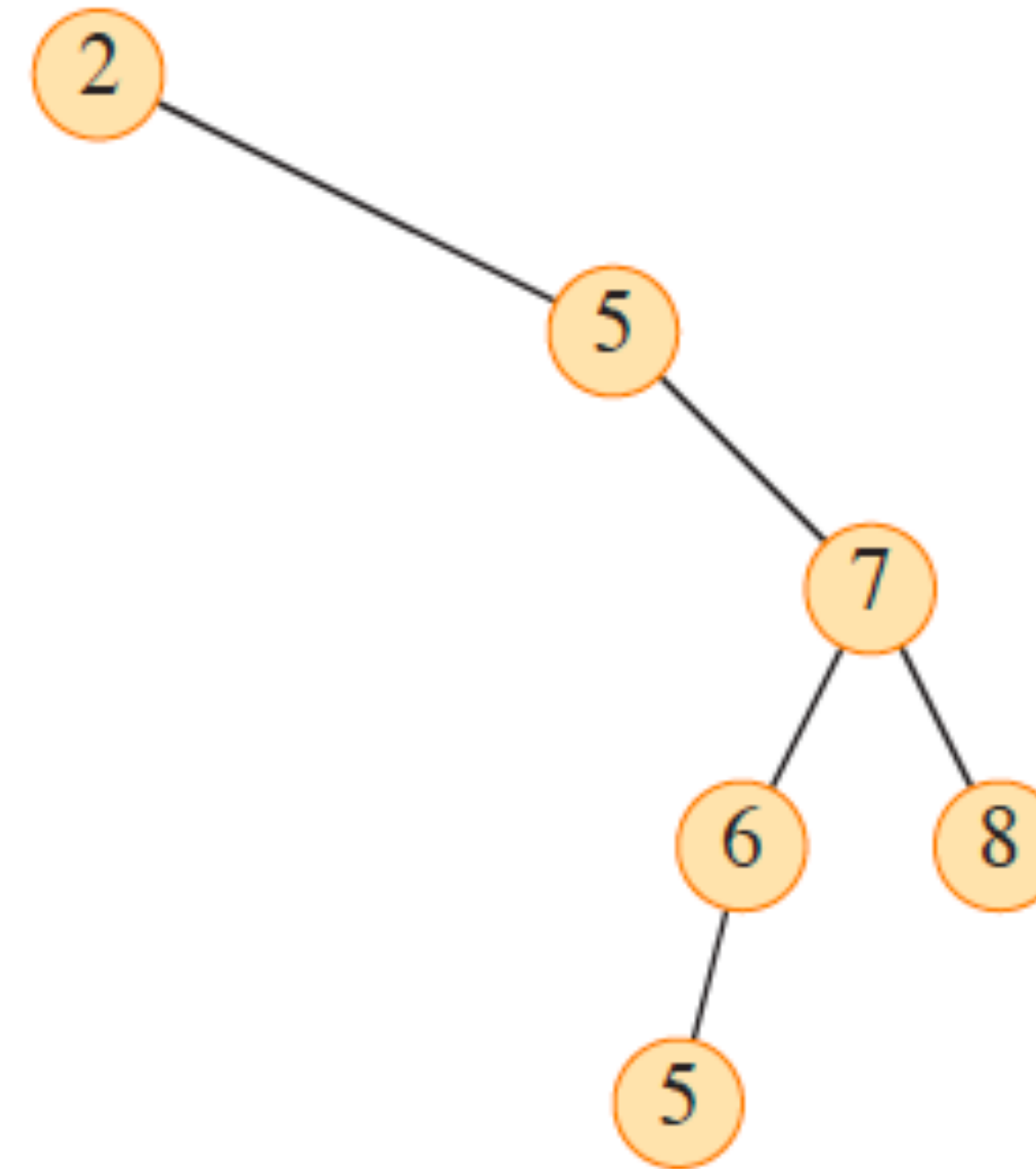
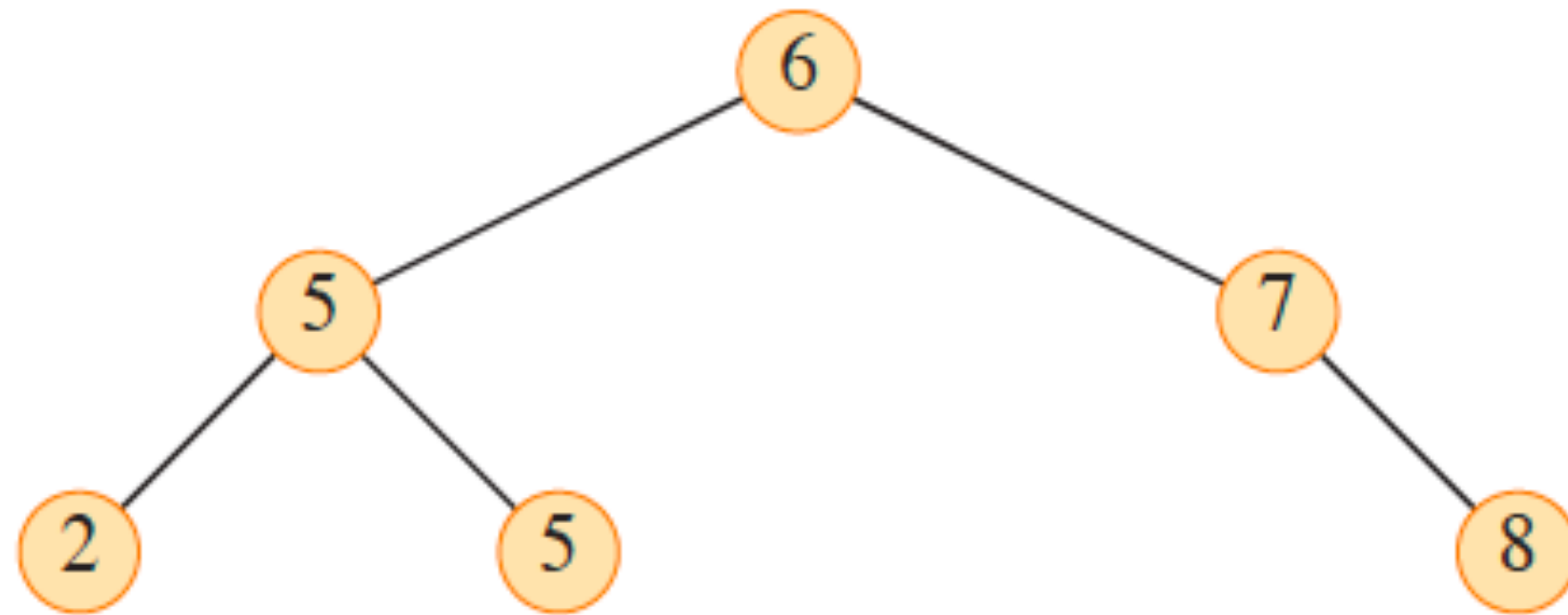
- **Insert** an element.
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- **Minimum** or **Maximum** of the set.
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What is a Binary Search Tree?

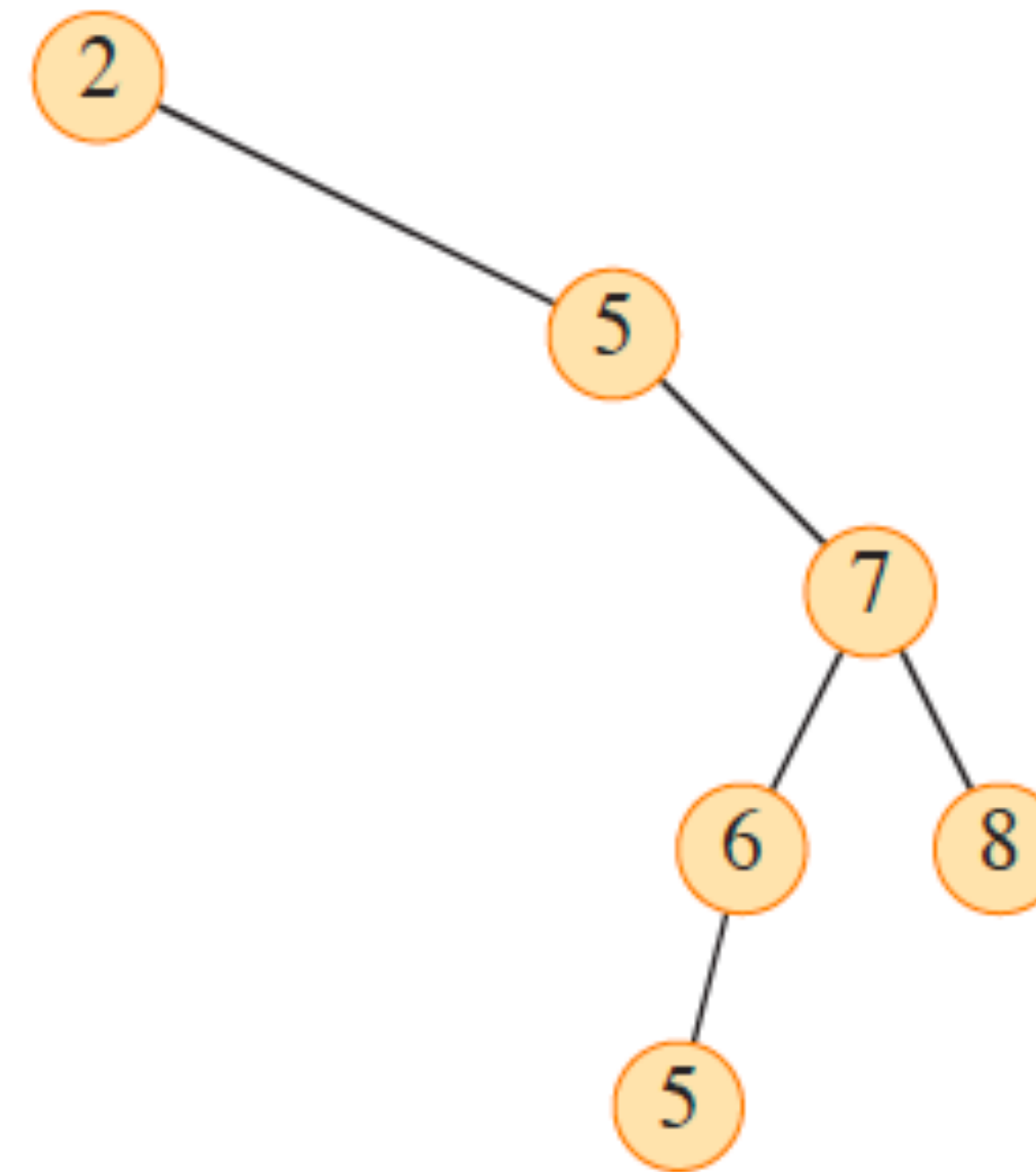
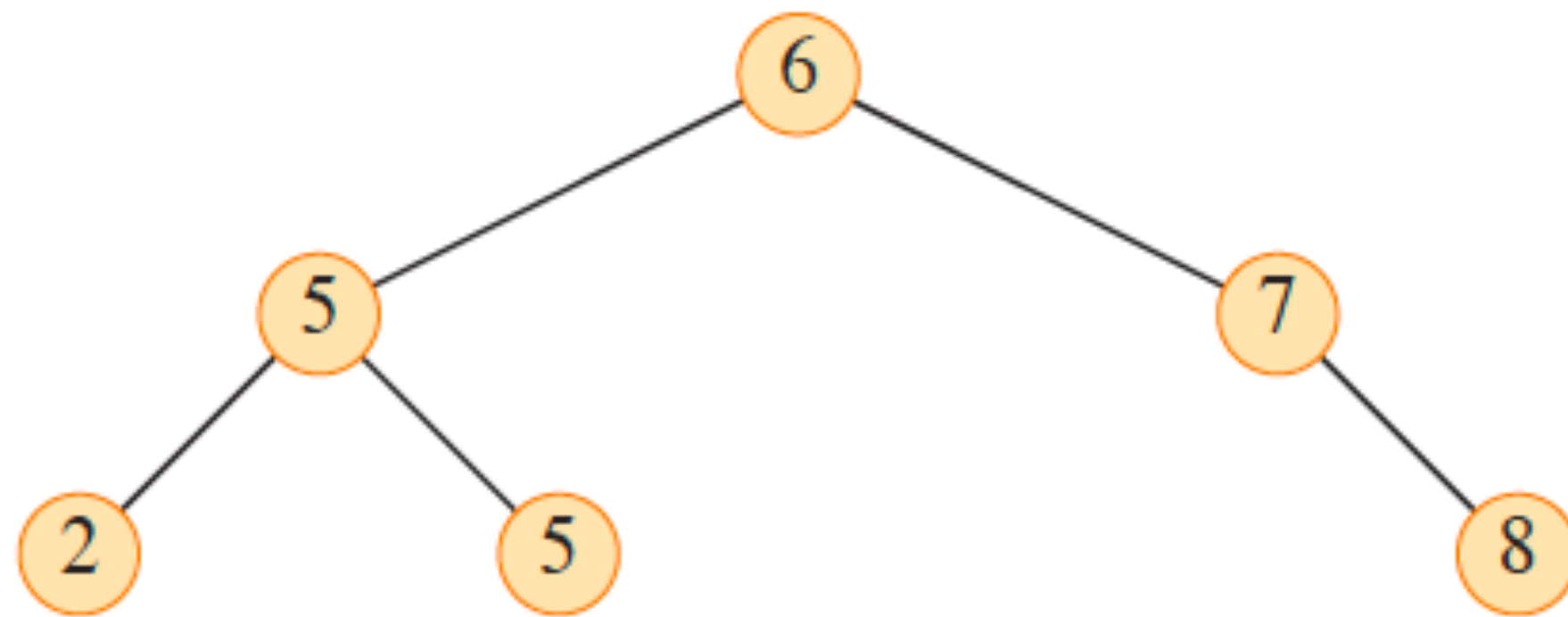
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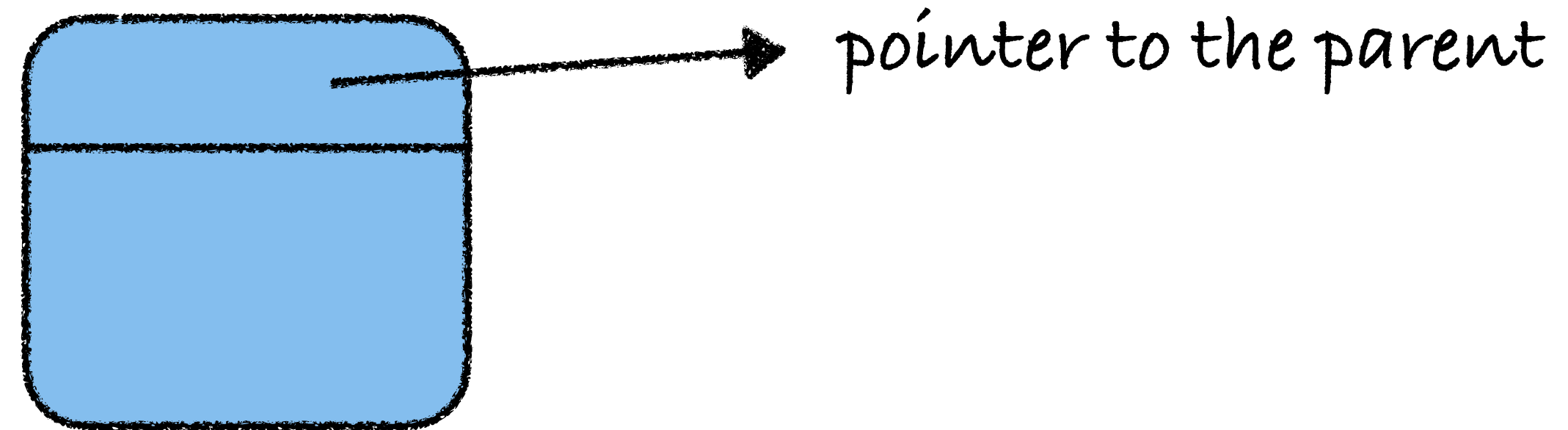
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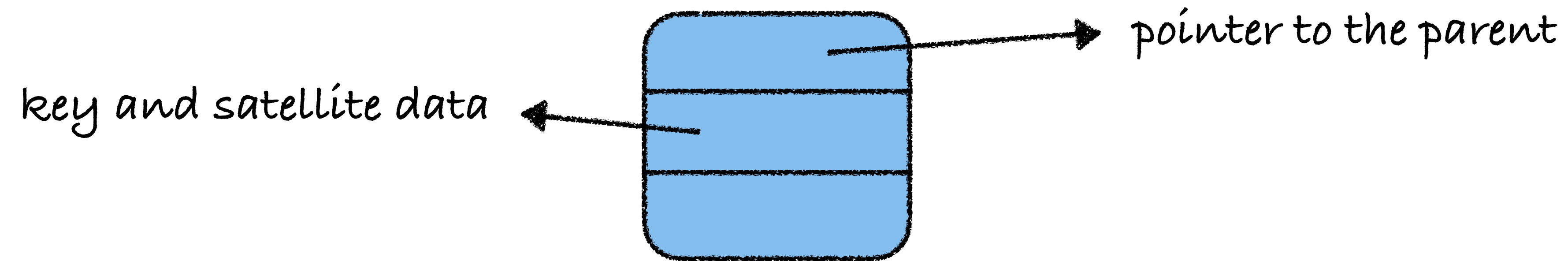
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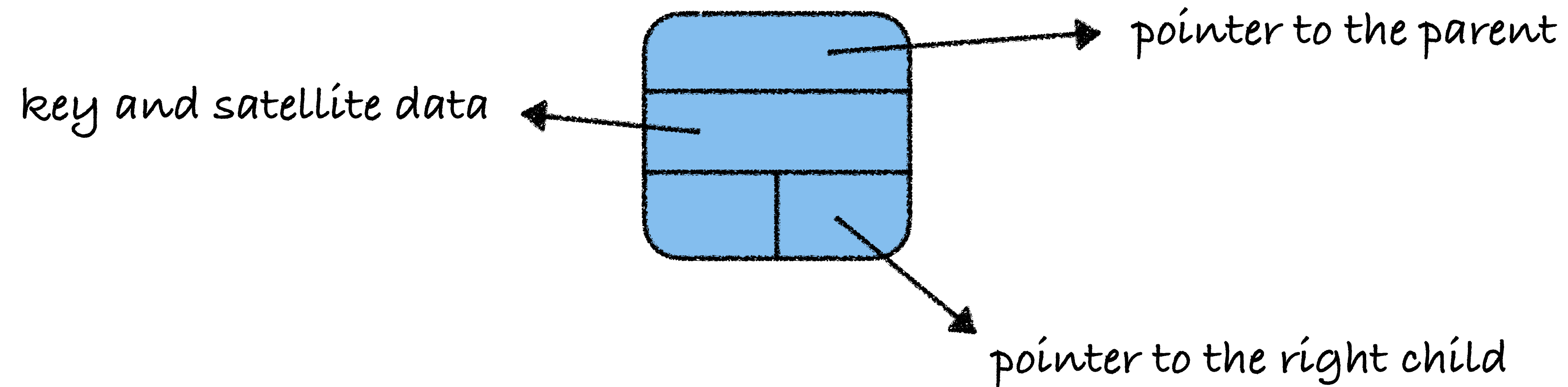
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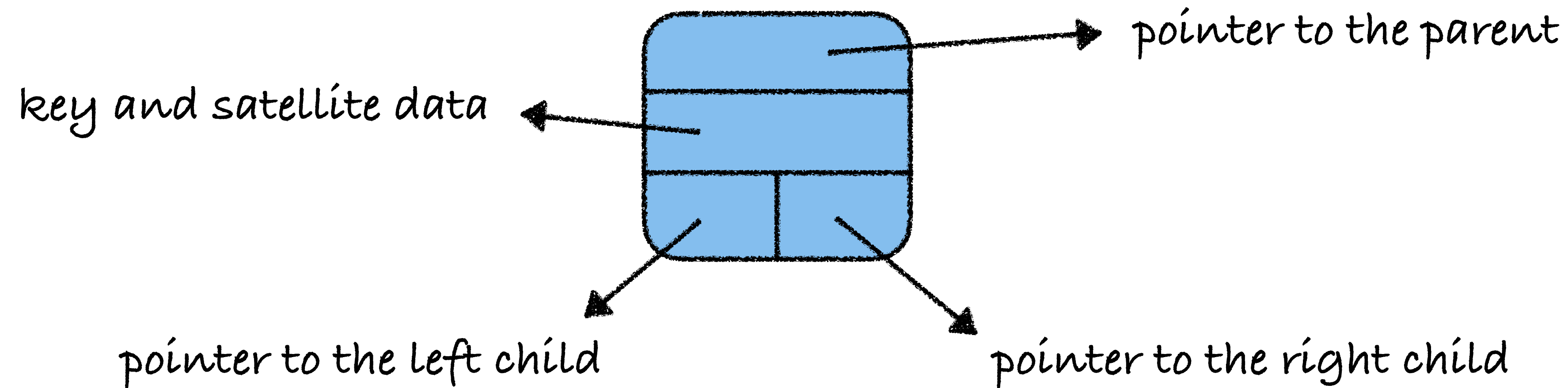
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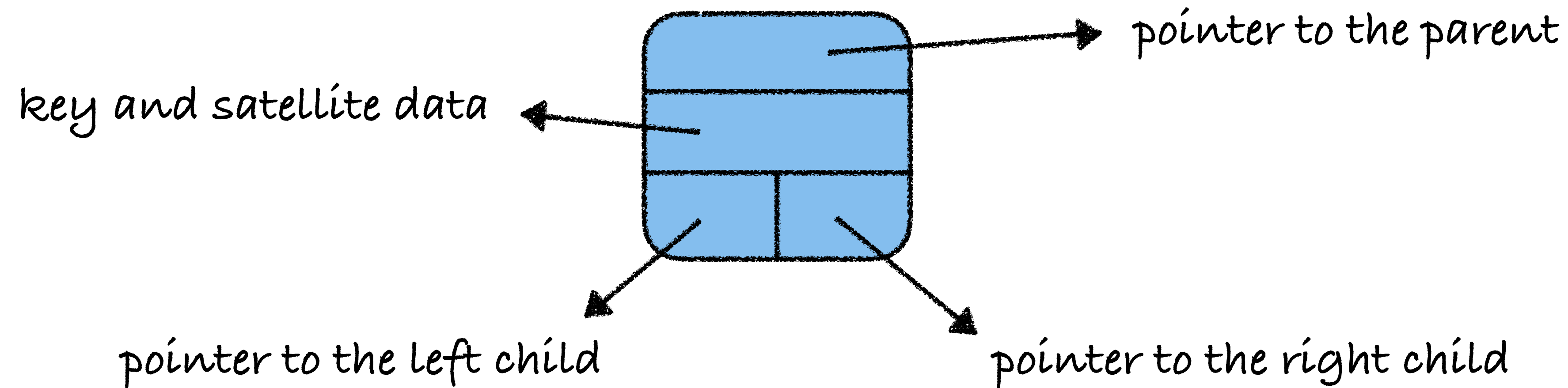
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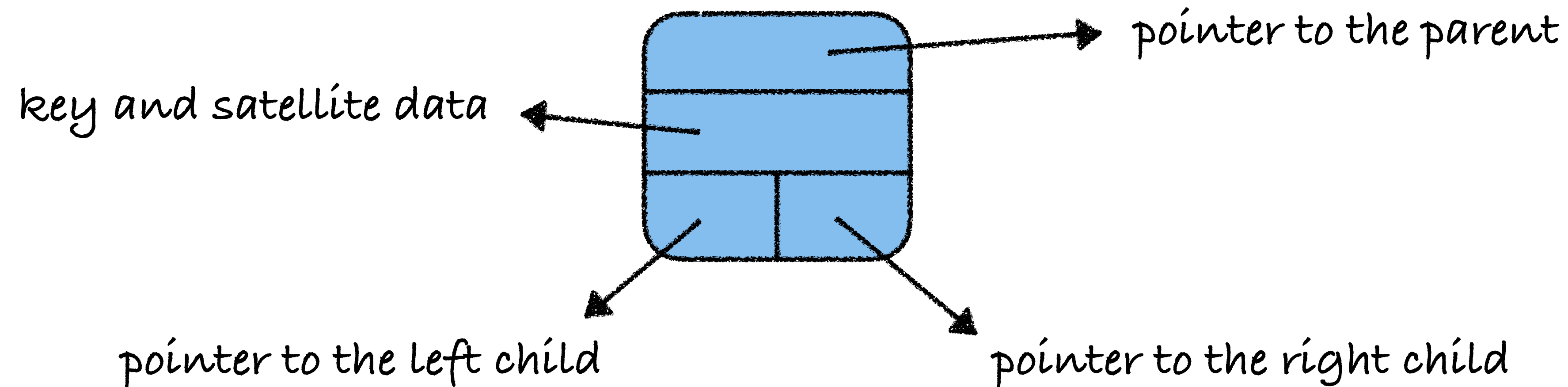
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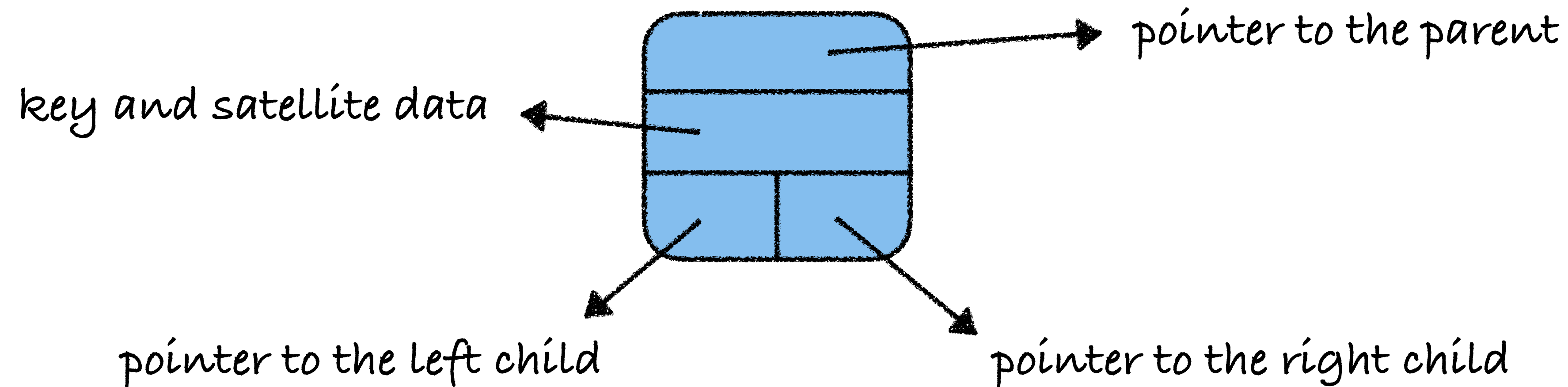
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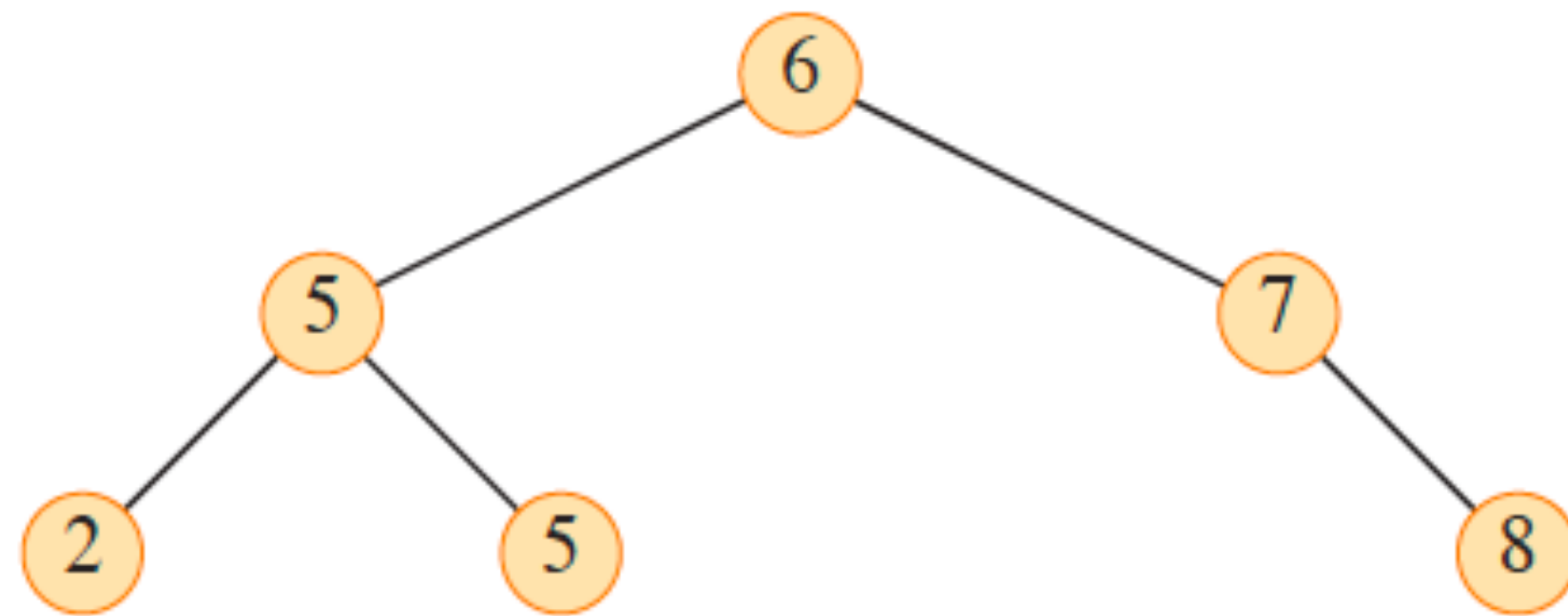
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Let x be a node in BST and y, z be the nodes in its left, right subtree, respectively.

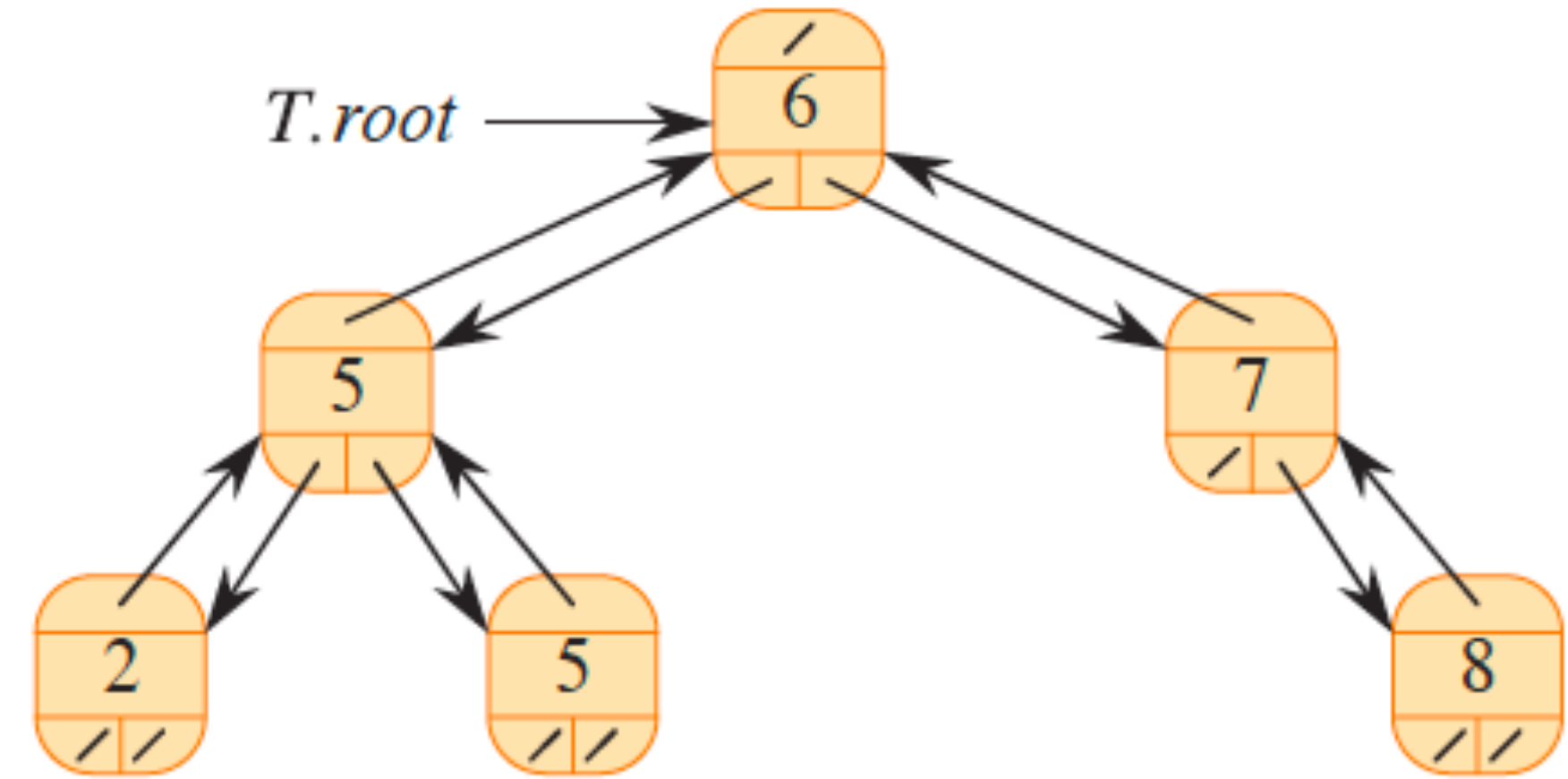
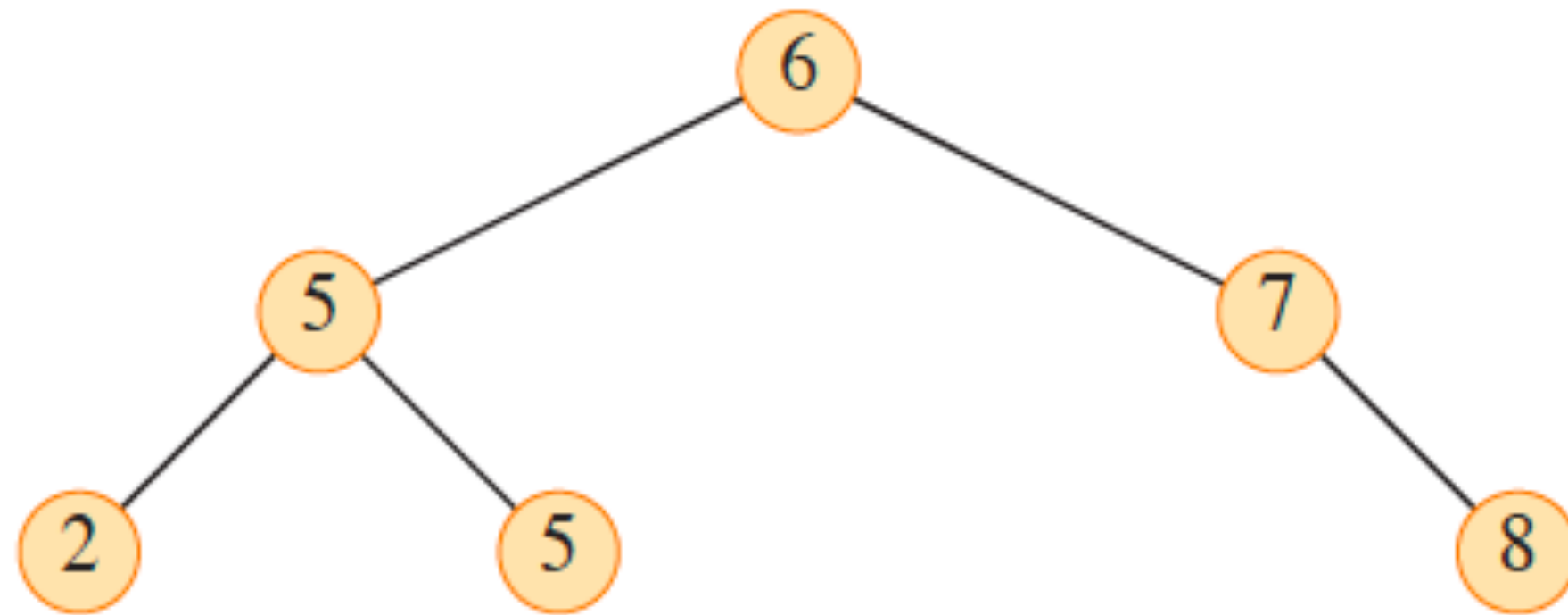
Then, $y.key \leq x.key \leq z.key$.

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