

# 01 – Trees 1

Advanced Algorithms and Data Structures



UNIVERSITY OF ZAGREB  
Faculty of Electrical  
Engineering and  
Computing

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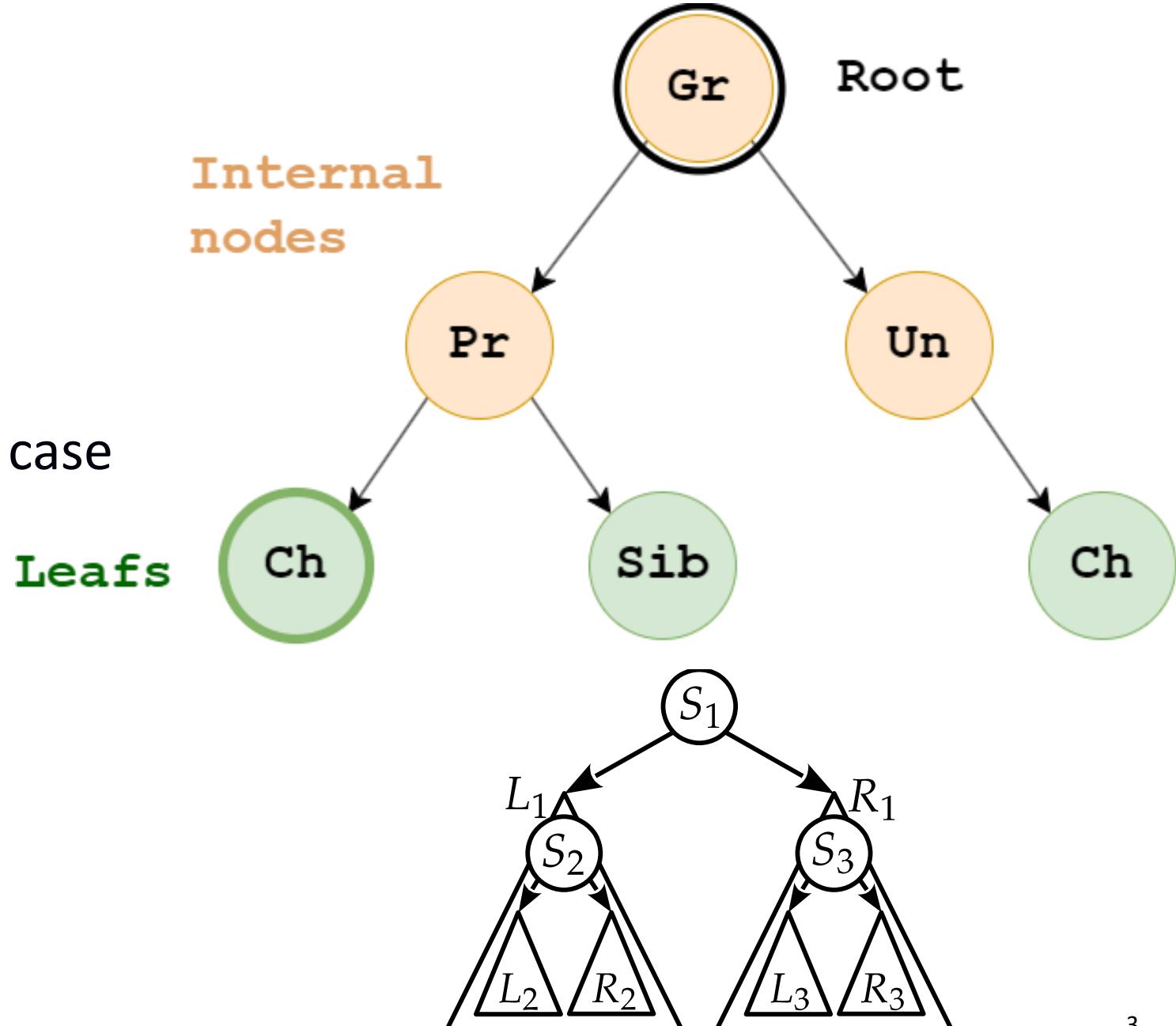


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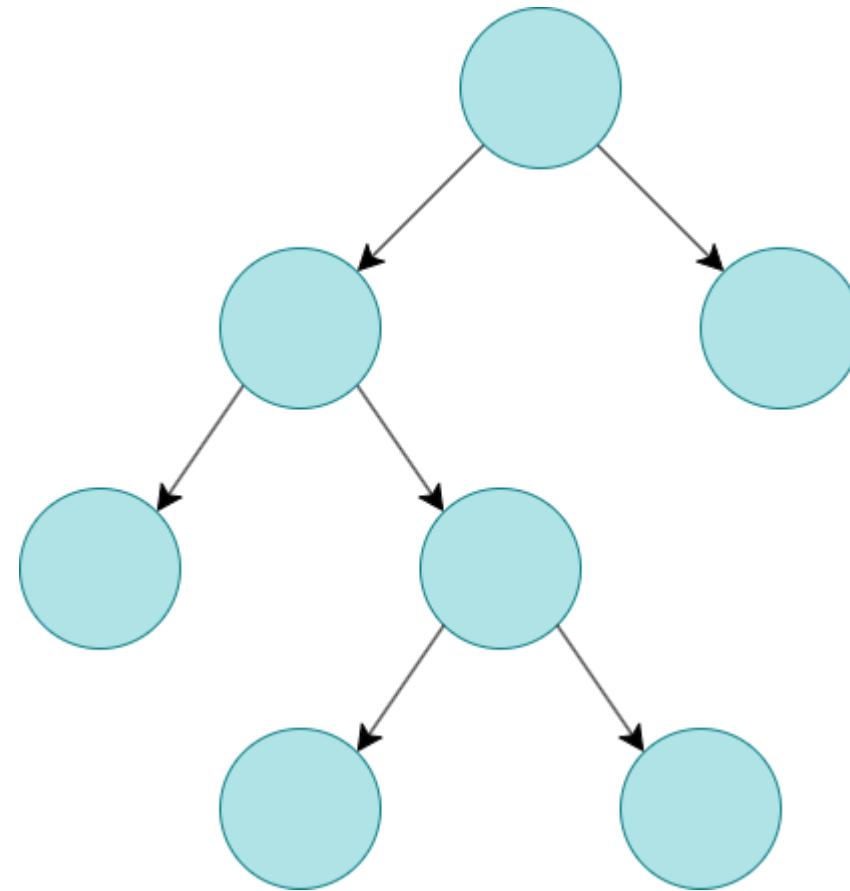
# Binary trees (1)

- Trees are **acyclic graphs**
- **Binary trees** are a special case
  - Maximum of 2 children
- Recursive definition:  
$$B = (L, S, R)$$



# Binary trees (2)

- **Full binary tree**
  - Nodes either have 0 or 2 children
  - *Internal nodes have 2 children*



# Binary trees (3)

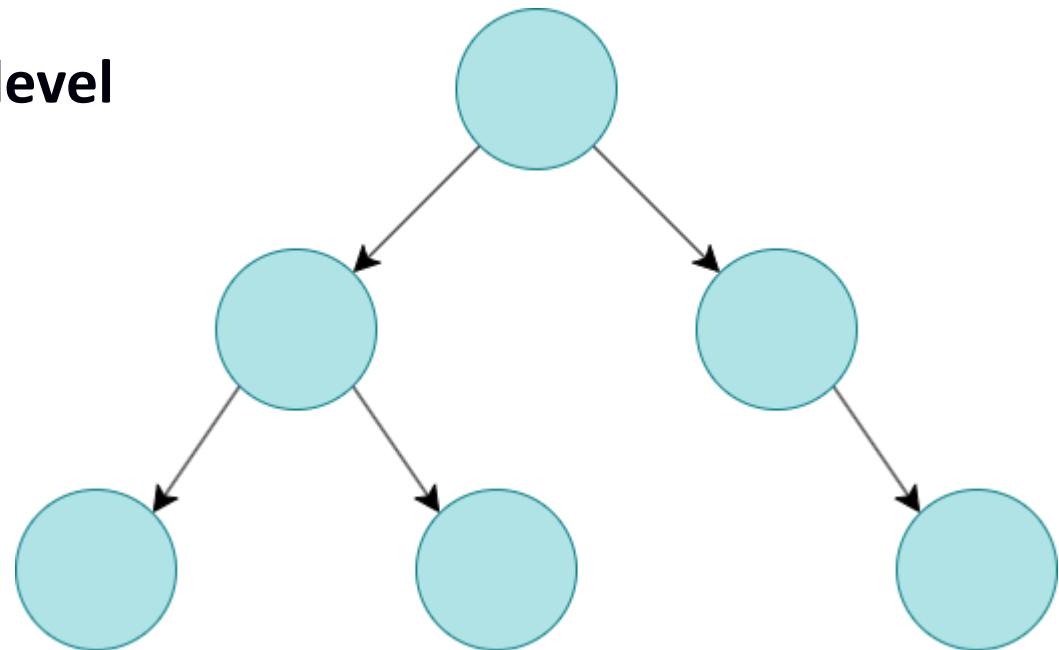
- **Complete** binary tree
  - All levels filled, **except for the lowest level**

$$N_{nodes} \leq 2^h - 1$$

$$N_{leafs} (h=level) \leq 2^{h-1}$$

$$N_{internal\ nodes} \leq 2^{h-1} - 1$$

$$H = \lceil \log_2(n + 1) \rceil$$



# Binary trees (4)

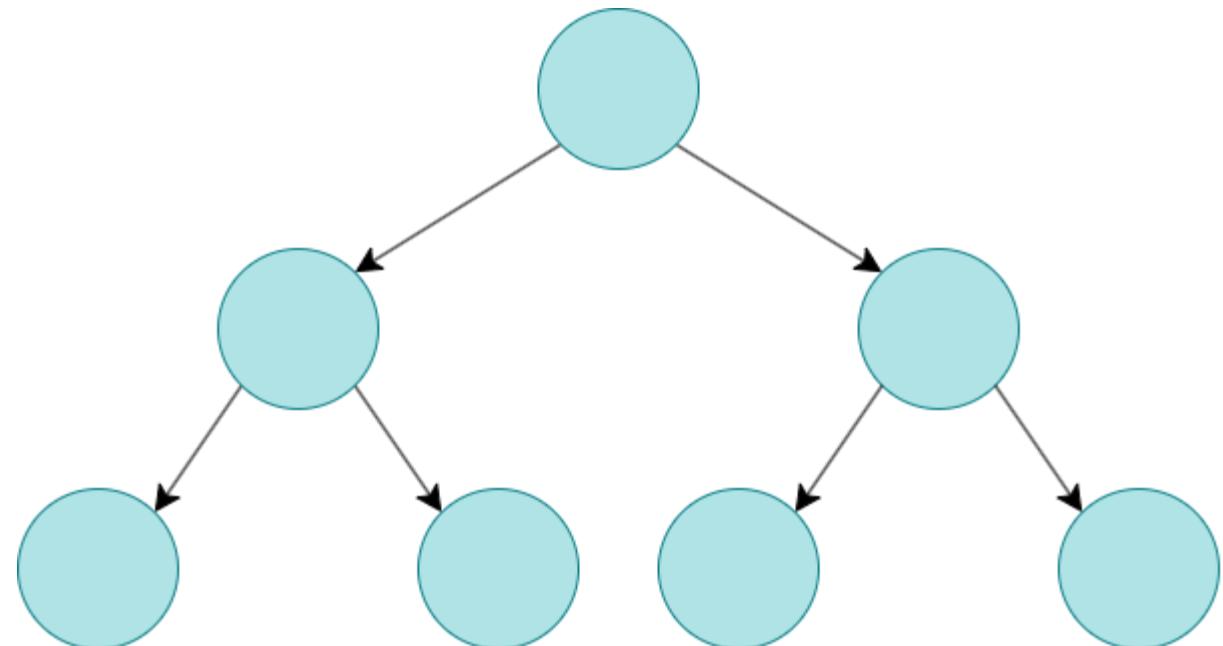
- **Perfect** binary tree
  - All levels are full
- Important formulae:

$$N_{nodes} = 2^h - 1$$

$$N_{leafs} (h=level) = 2^{h-1}$$

$$N_{internal\ nodes} = 2^{h-1} - 1$$

$$H = \log_2(n + 1)$$



# Binary tree (5)

- **Sorted** binary tree

$$B = (L, S, R)$$

- No duplicates:

$$\text{val}(S(L)) < \text{val}(S) < \text{val}(S(R))$$

- With duplicates:

$$\text{val}(S(L)) \leq \text{val}(S) < \text{val}(S(R))$$

$$\text{val}(S(L)) < \text{val}(S) \leq \text{val}(S(R))$$

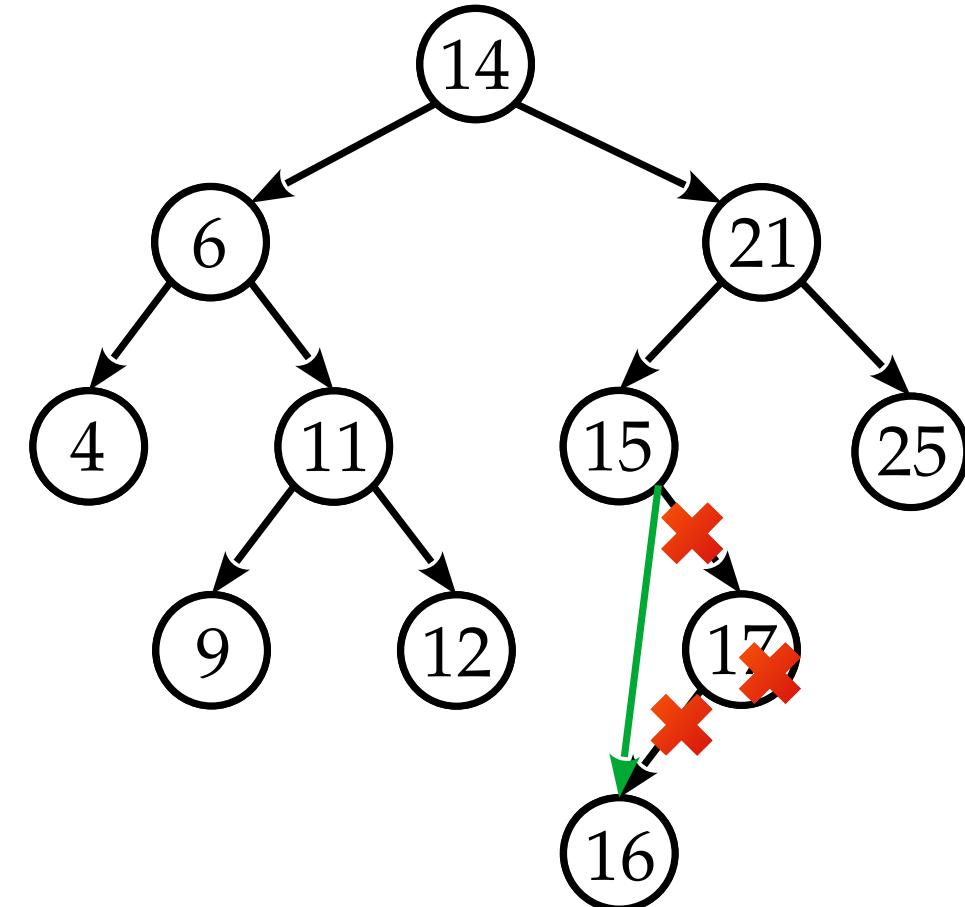
- Lesser valued nodes go LEFT
- Greater valued nodes go RIGHT

# Operations - adding a node (SBT)

- Search from root to leaves
  - Left if lesser
  - Right if greater
- At leaf, place node:
  - As left child if lesser
  - As right child if greater

# Operations - deleting a node (SBT)

- Search from root to node
  - ...
- **Sc. 1** The node **is a leaf** – REMOVE
- **Sc. 2** The node **has 1 child**
  - Remove node
  - Recombine Gr -> Ch
  - Side is kept according to Gr „pointer”
- 2 children??

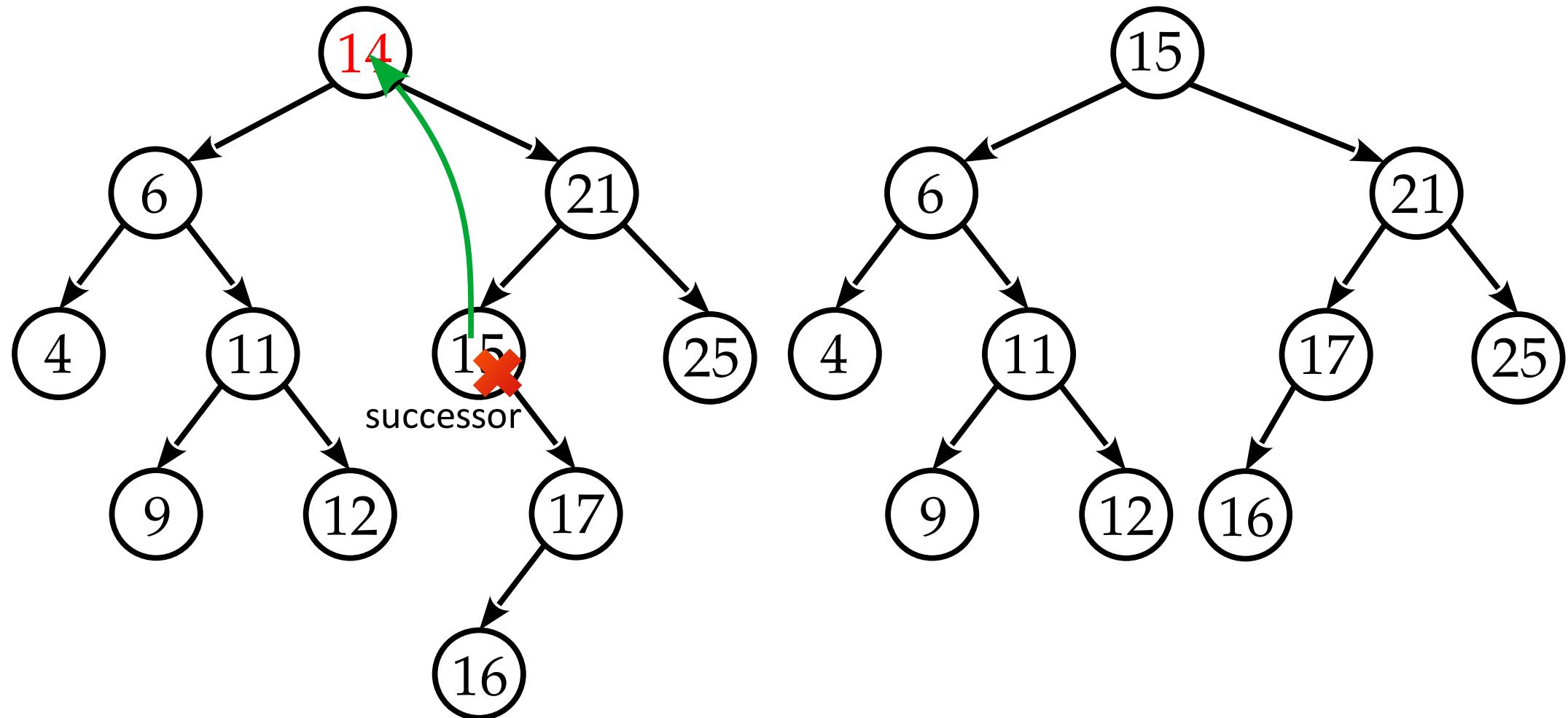


# Operations – Deleting by copying (1)

- **Strategy 1: By copying**
  - Structural changes are minimal
  - Comes down to Sc. 1 and 2
- Find the rightmost node in the left subtree – predecessor
  - or ...
- Find the leftmost node in the right subtree – successor
- The ...essors have 0 or 1 child

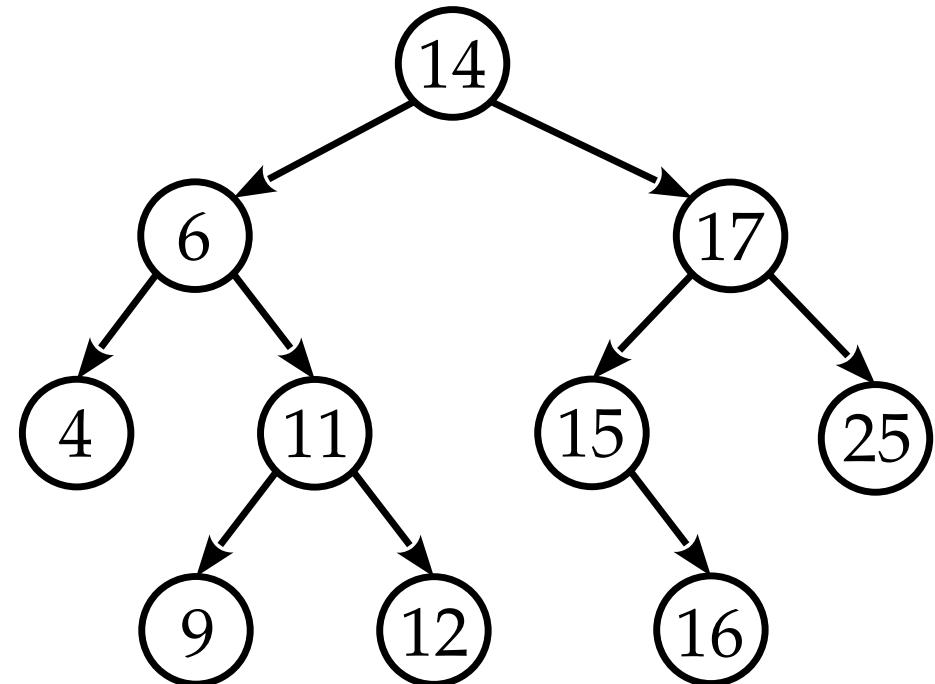
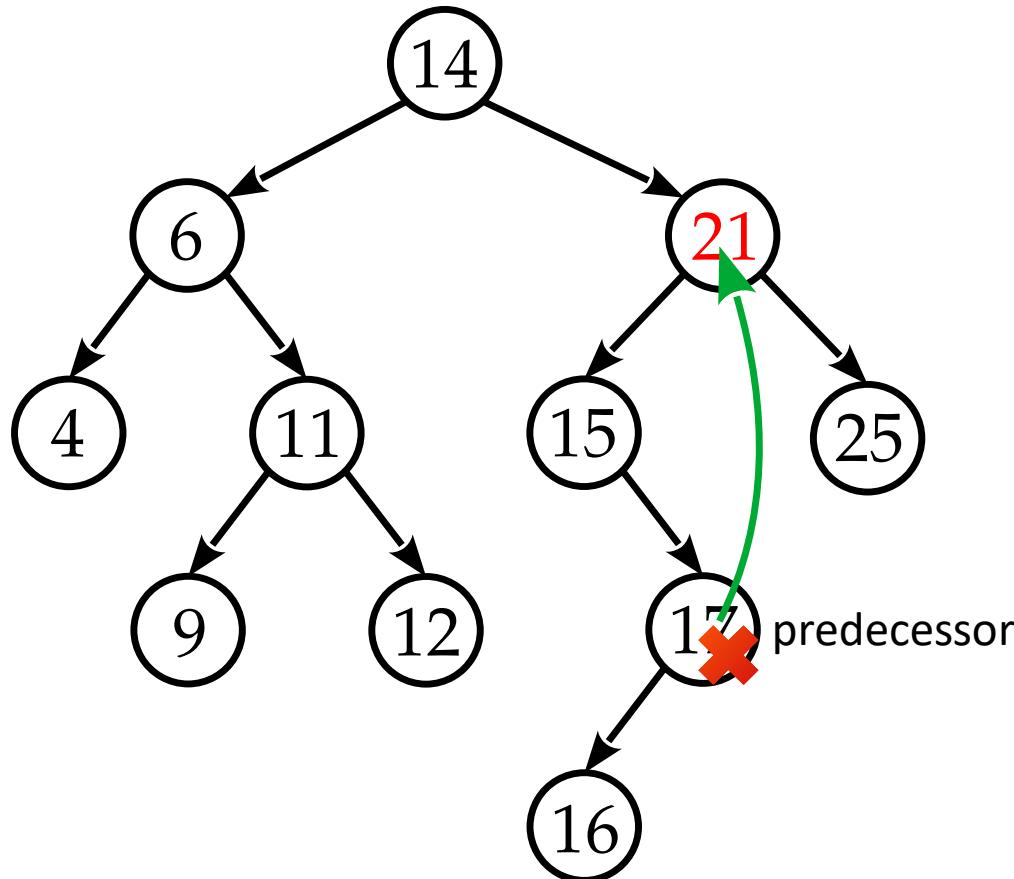
# Operations – Deleting by copying (2)

- Delete 14



# Operations – Deleting by copying (3)

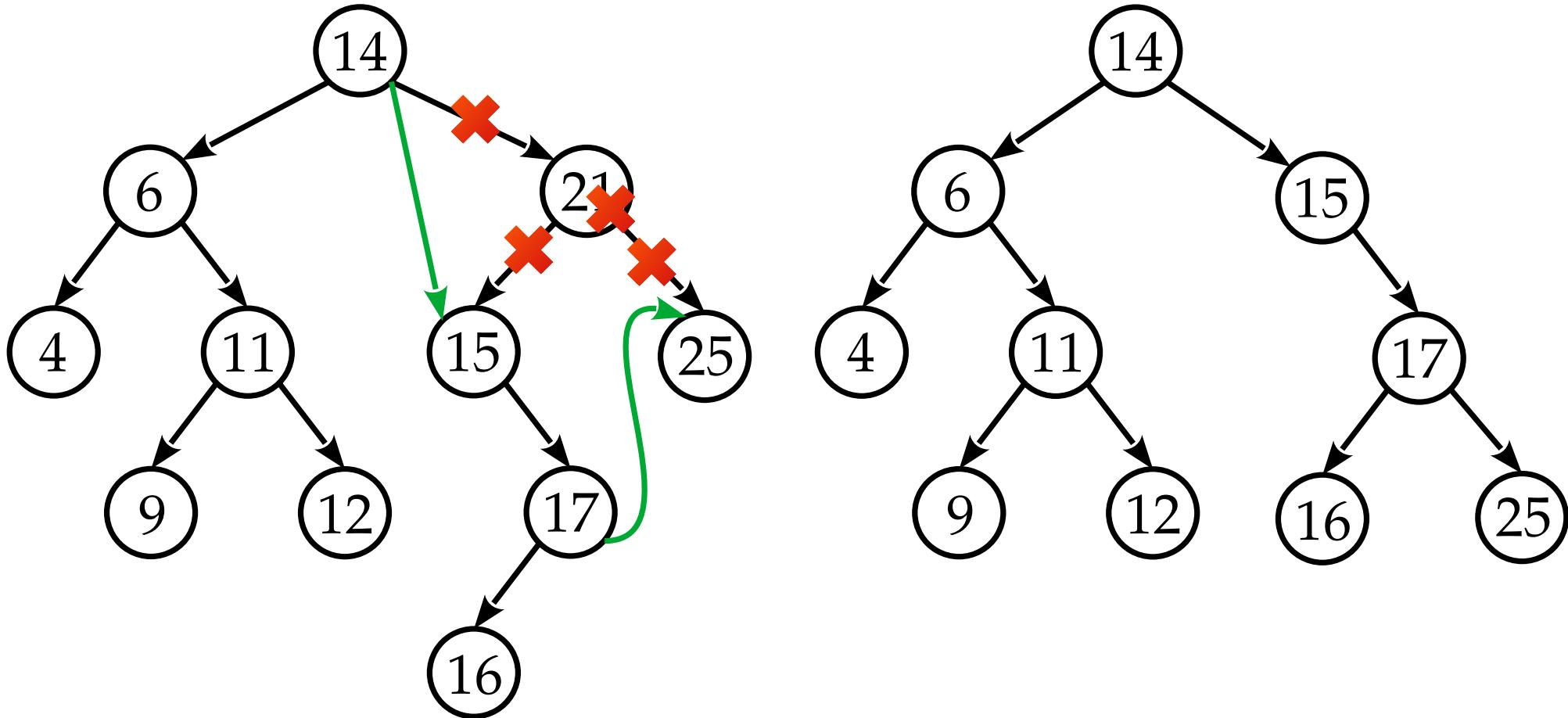
- Delete 21



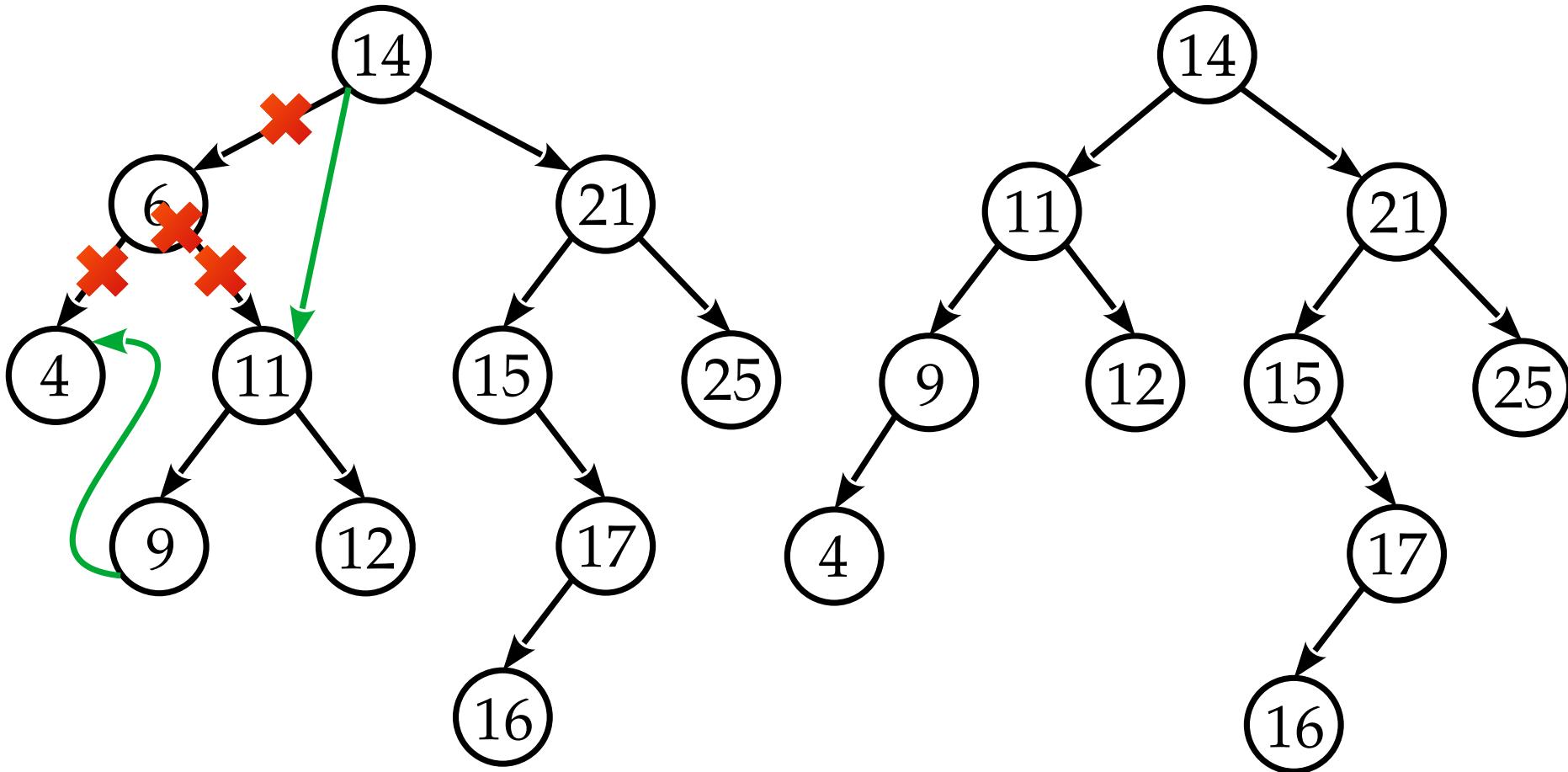
# Operations - Deleting by merging (1)

- **Strategy 2: By merging**
  - Structural changes are considerable
  - The idea is to merge the subtrees of a node
- Find the *target* and *Pr*
  - Determine the side of the *target* in relation to the *Pr*
  - If left subtree – the merge subtree is *target->right*
    - *target-> left* goes to the leftmost of the merge subtree
  - If right subtree – the merge subtree is *target->left*
    - *target -> right* goes to the rightmost of the merge subtree

# Operations - Deleting by merging (2)

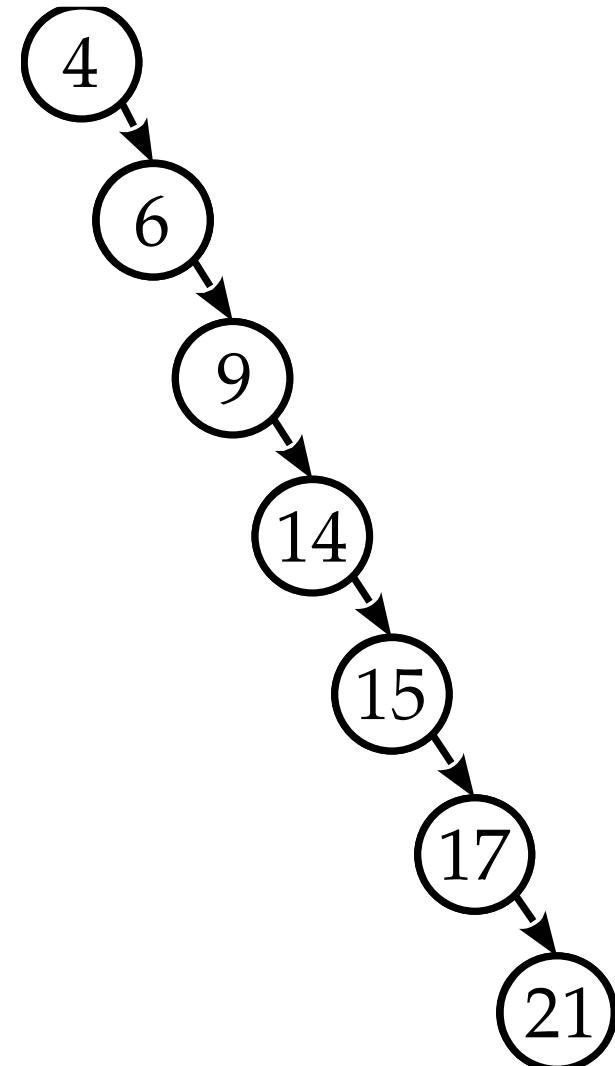


# Operations - Deleting by merging (3)



# Outcome of operations?

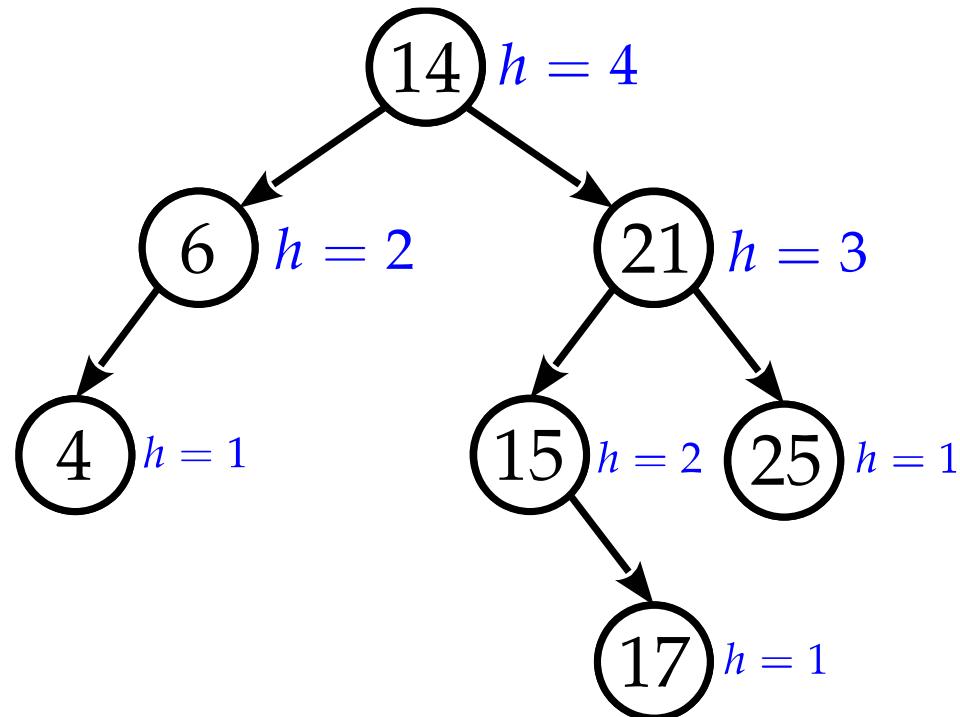
- Degenerate tree (backbone)
  - Left or right skewed
  - Spine
  - Linked list
- Search complexity (time)?
- How to fix this?



# Balanced tree

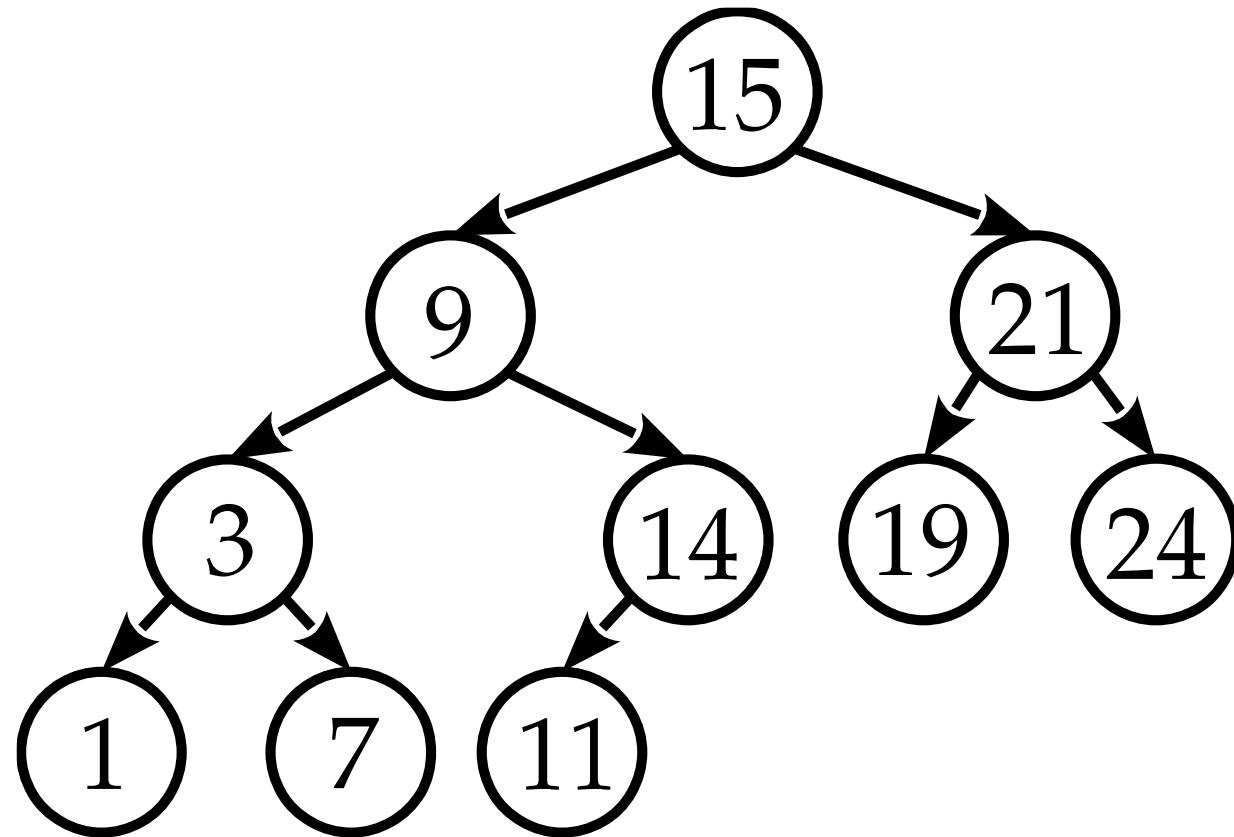
- For a binary tree  $B = (L, S, R)$  a balanced tree has the property  
 $\forall S \quad |h(L) - h(R)| \leq 1$
- ALL left and right subtrees have to be approximately the same height

- Perfectly balanced tree



# Balanced tree?

- Minimum deletions to disbalance?

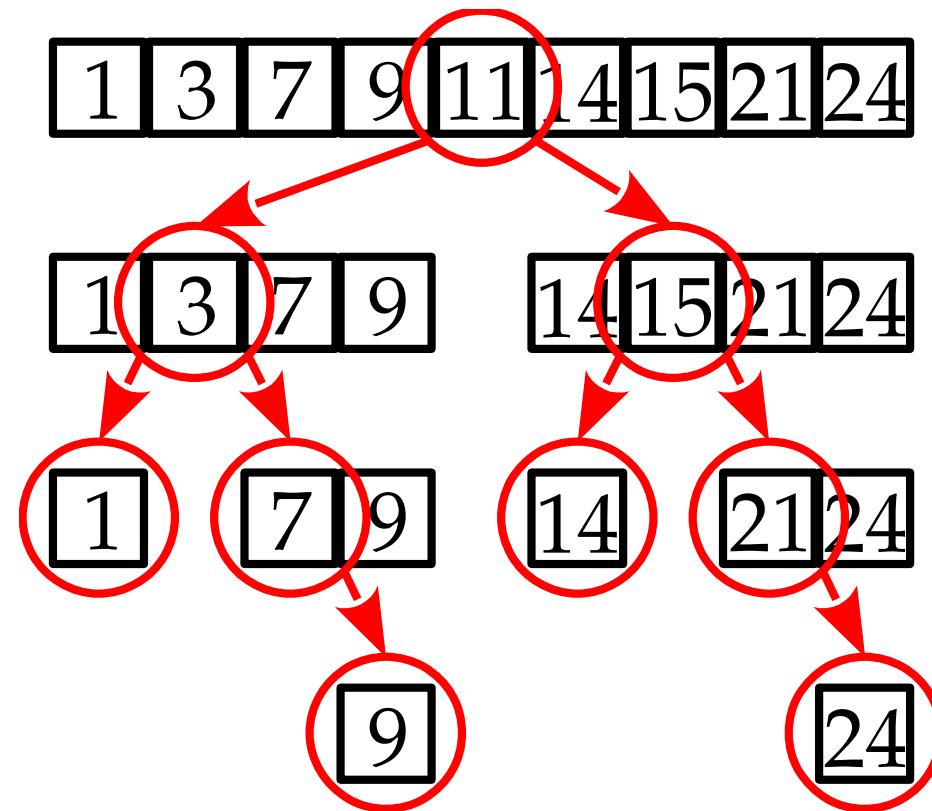


# Creating a balanced tree (sorted array)

- Get an array of values
- Sort the array
- Split the array
  - Positional centers become roots

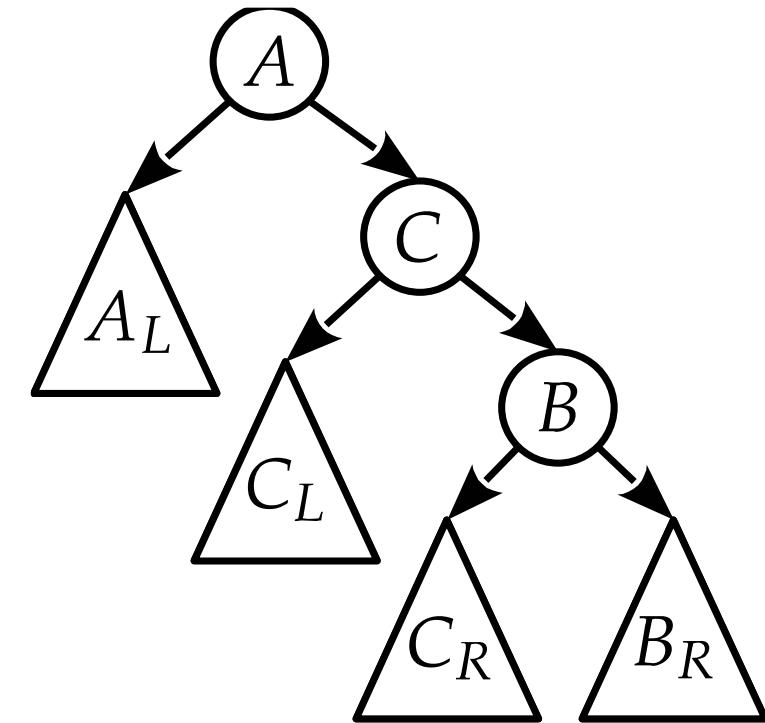
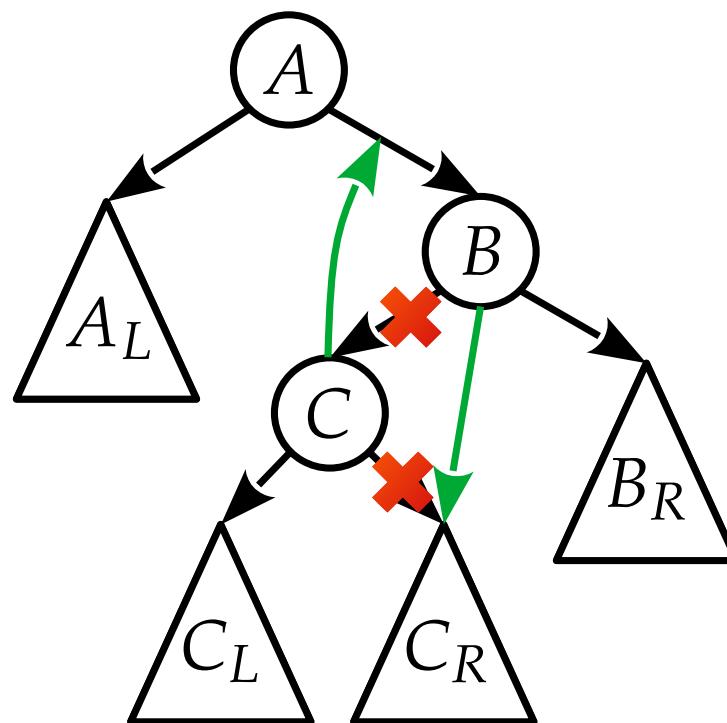
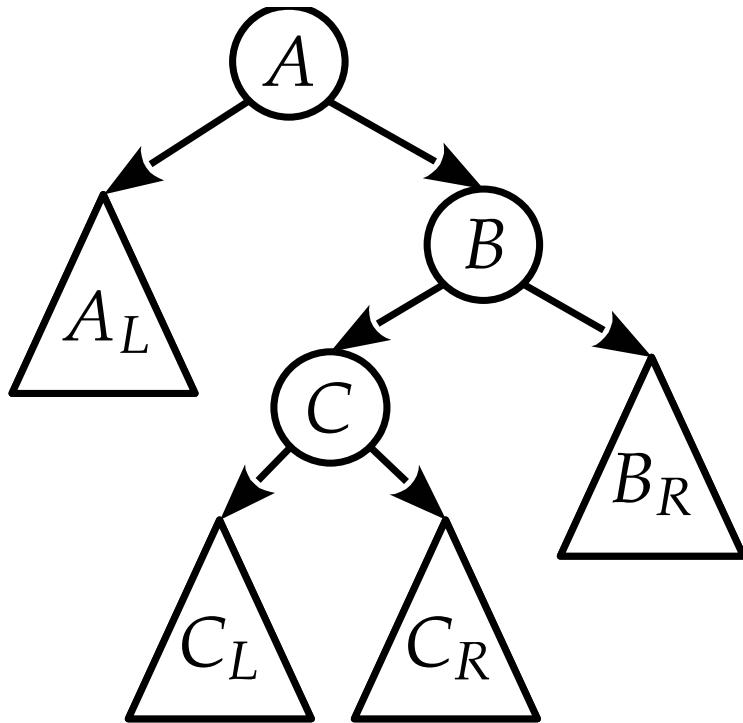
$O(n \log_2 n + n)$   
sort + pass through array

$$V_{sorted} = [1, 3, 7, 9, 11, 14, 15, 21, 24]$$



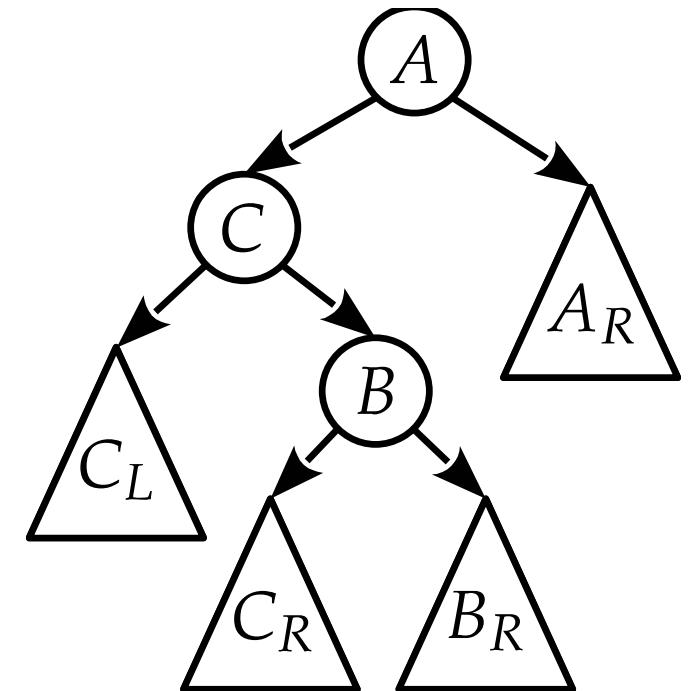
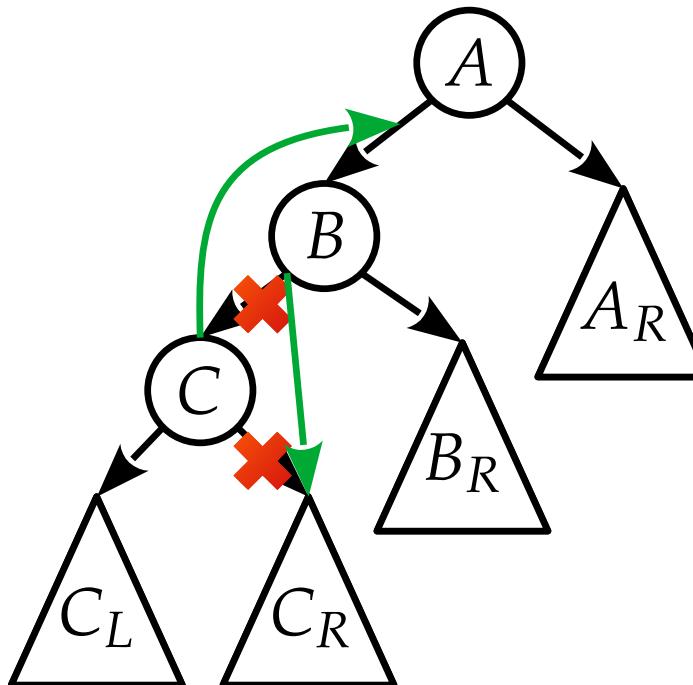
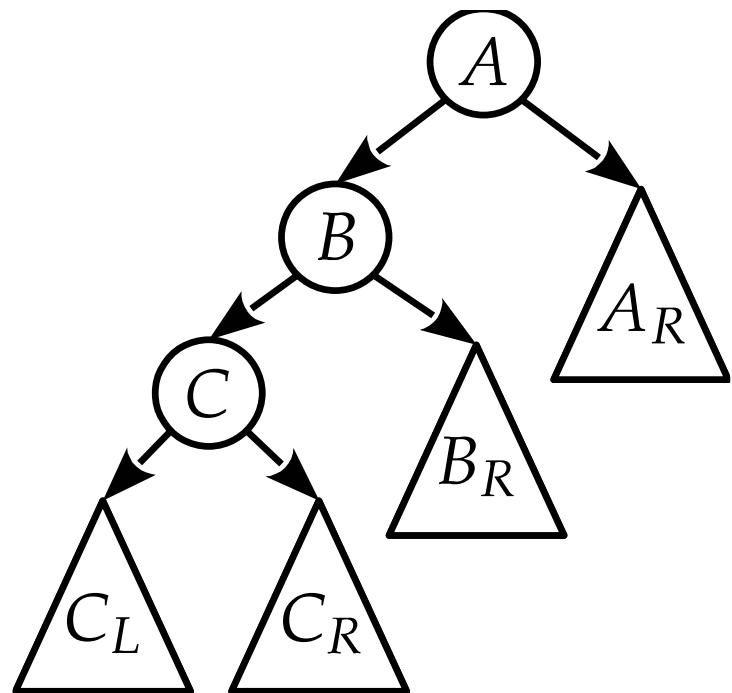
# Rotations – right (1)

- C over B
- Get B as C->right



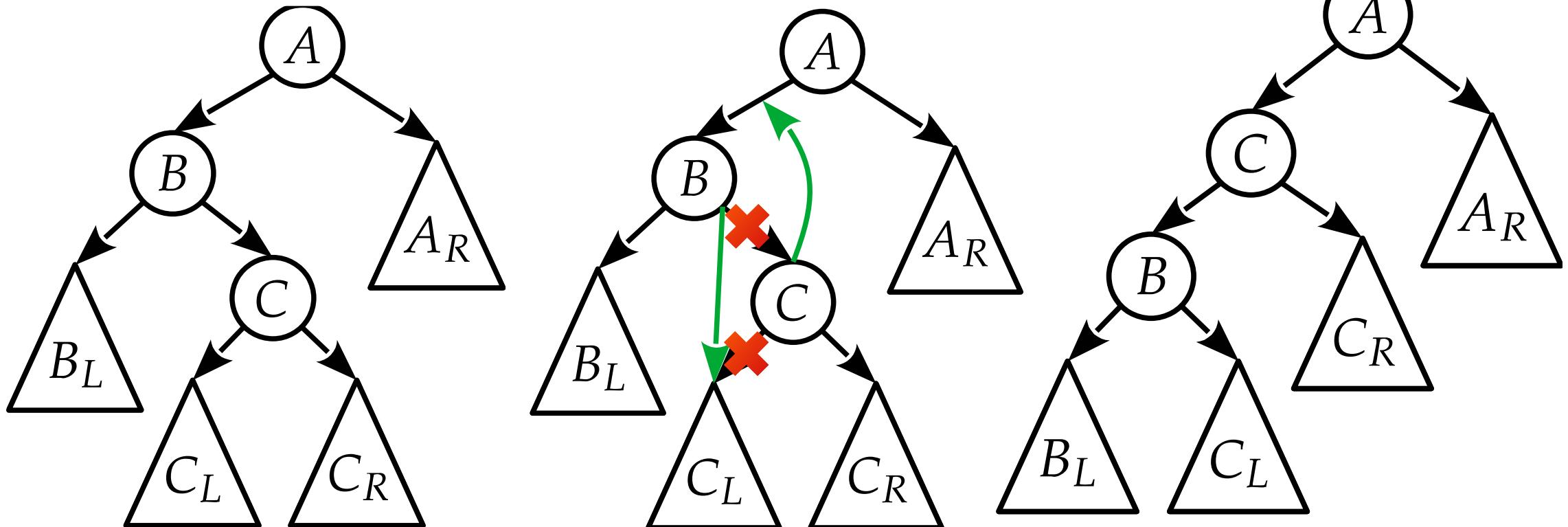
# Rotations – right (2)

- C over B
- Get B as C->right



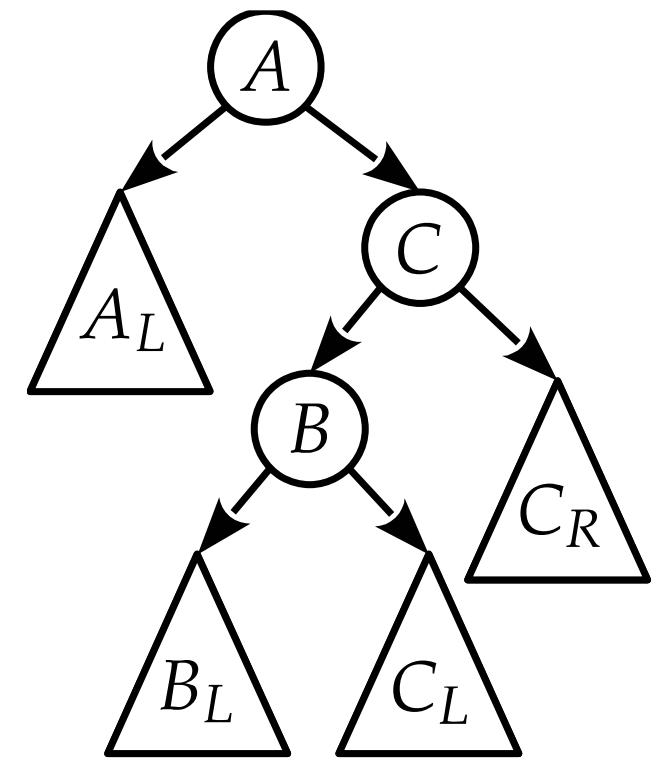
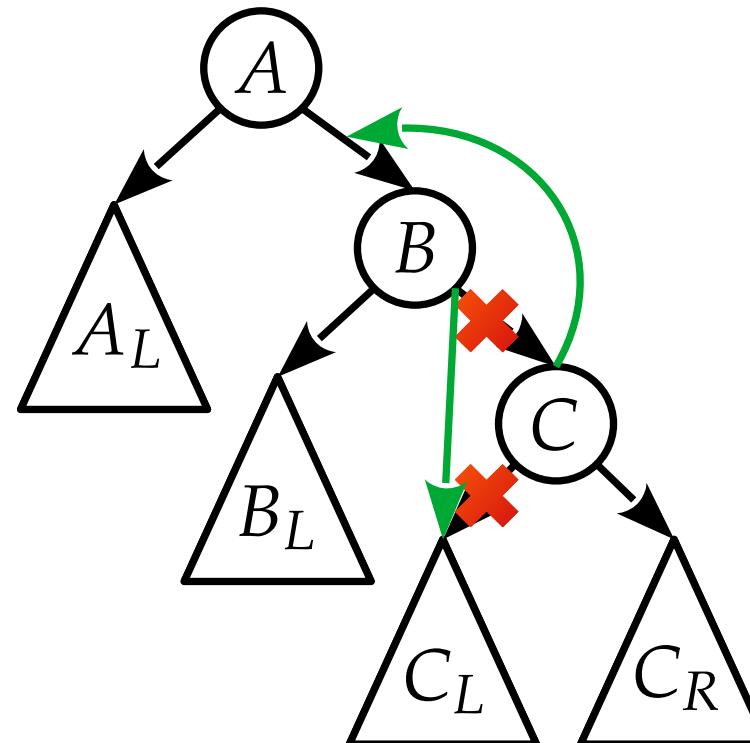
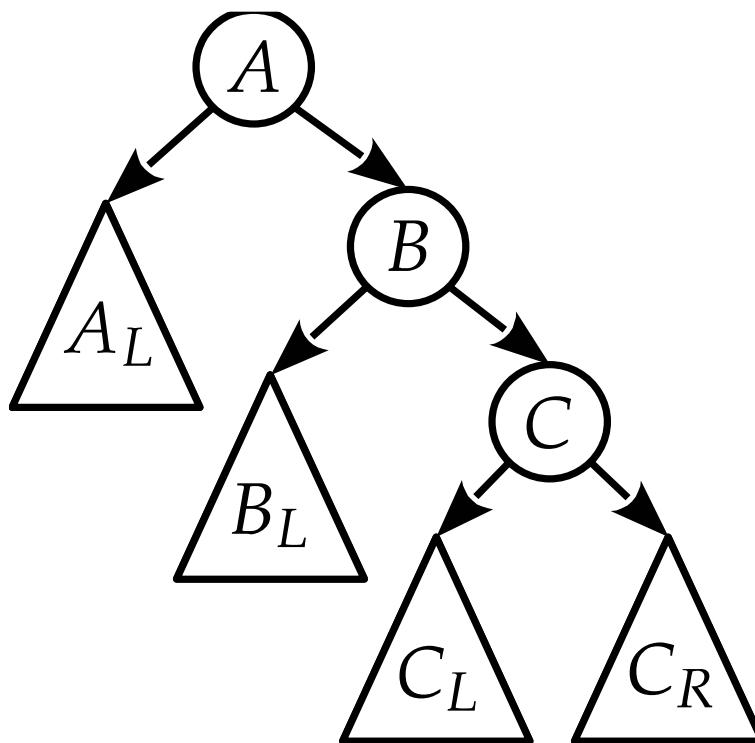
# Rotations – left (1)

- C over B
- Get B as C->left



# Rotations – left (2)

- C over B
- Get B as C->left



# Day-Stout-Warren algoritam (DSW)

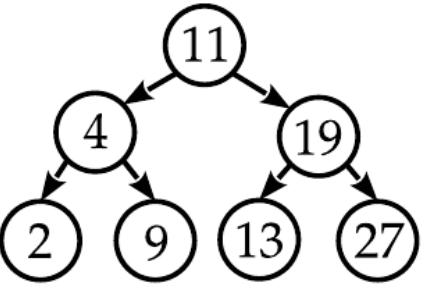
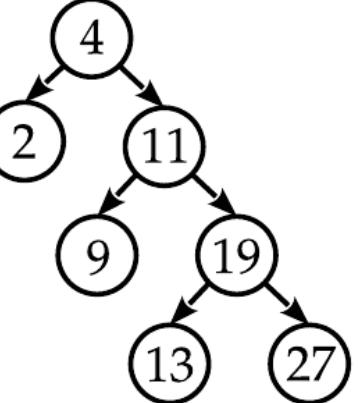
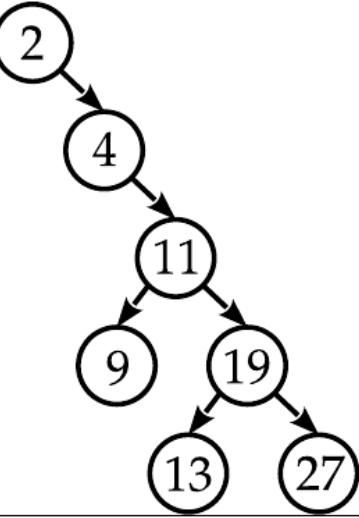
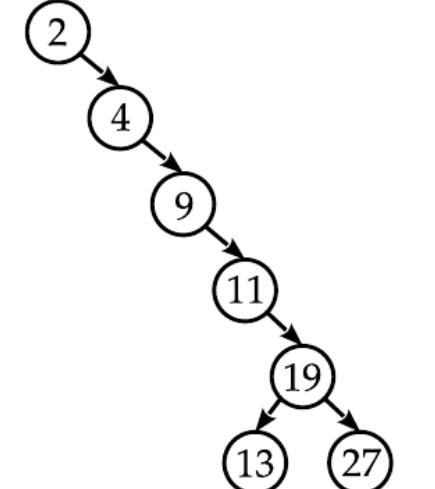
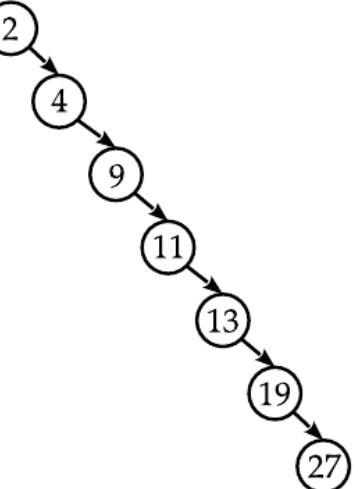
1. Create a backbone
  - Continuous rotations to L or R
2. Recursively break the backbone to get a complete tree

# DSW – backbone creation (1)

- Depending on the desired backbone:
  - Left children are rotated right
  - Right children are rotated left
- Repeat until backbone
  - No left children
  - No right children
- RightRotate (C, B)
  - Missing something?

```
procedure RIGHTBACKBONE(root)
    B  $\leftarrow$  root
    A  $\leftarrow$  nil
    while B  $\neq$  nil do
        C  $\leftarrow$  leftChild(B)
        if C  $\neq$  nil then
            RIGHTROTATE(C, B)
            if A = nil then
                root  $\leftarrow$  C
                B  $\leftarrow$  C
            else
                 $\triangleright$  Descending right to the first node that has the left child
                A  $\leftarrow$  B
                B  $\leftarrow$  rightChild(B)
```

# DSW – backbone creation (2)

$A = \text{nil}, B = 11, C = 4$ 	$A = \text{nil}, B = 4, C = 2$ 	$A = 4, B = 11, C = 9$ 
$A = 11, B = 19, C = 13$ 	The final right backbone 	

# DSW – backbone breaking

- Strategically placed left rotations
- Complexity is  $O(n)$
- $n$  is known
- Determine number of nodes in the closest complete tree:

$$m = 2^{\lfloor \log_2(n+1) \rfloor} - 1$$

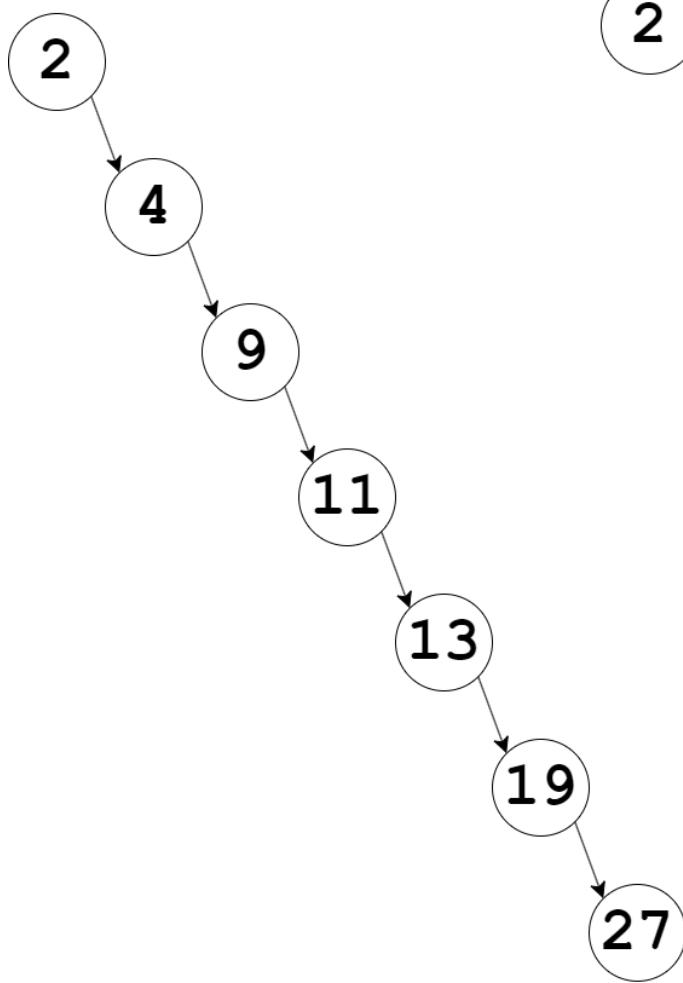
```
createPerfectTree(n) :  
    m = pow(2, floor(log2(n+1)))-1;  
    make n-m rotations from the top of the backbone; //overflowing nodes  
    while (m > 1):  
        m = m/2  
        make m rotations from the top of the backbone;
```

# DSW – backbone breaking example 1

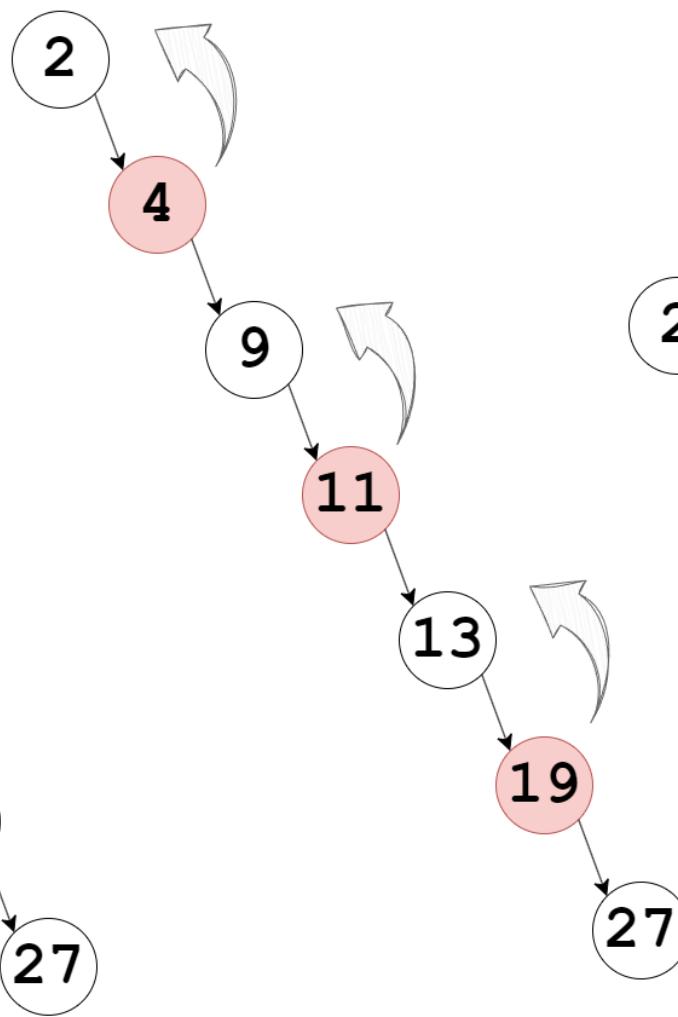
$$n = 7$$

$$m = 7$$

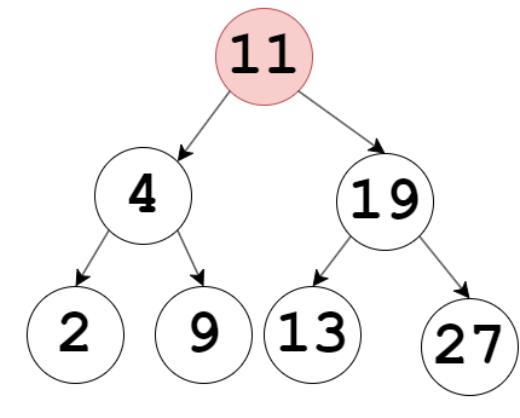
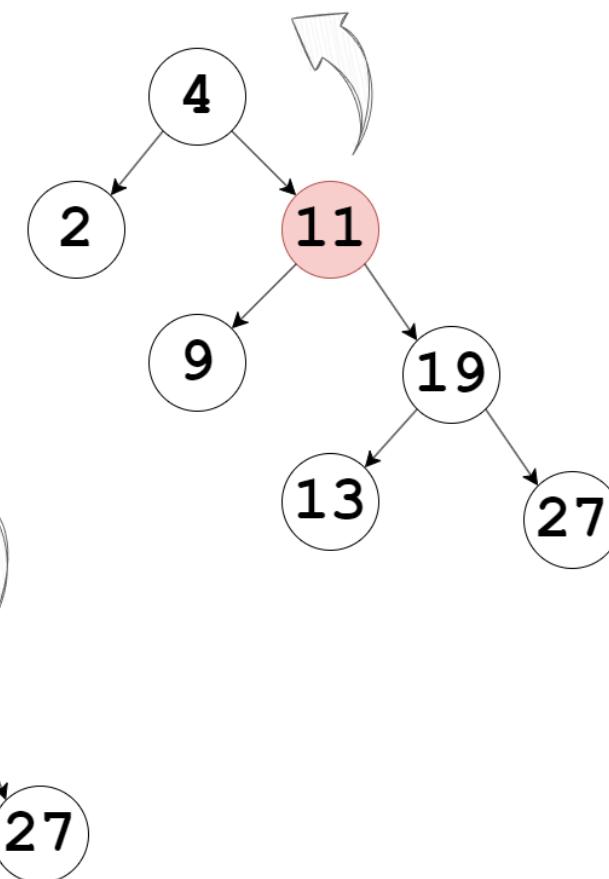
$$n - m = 0$$



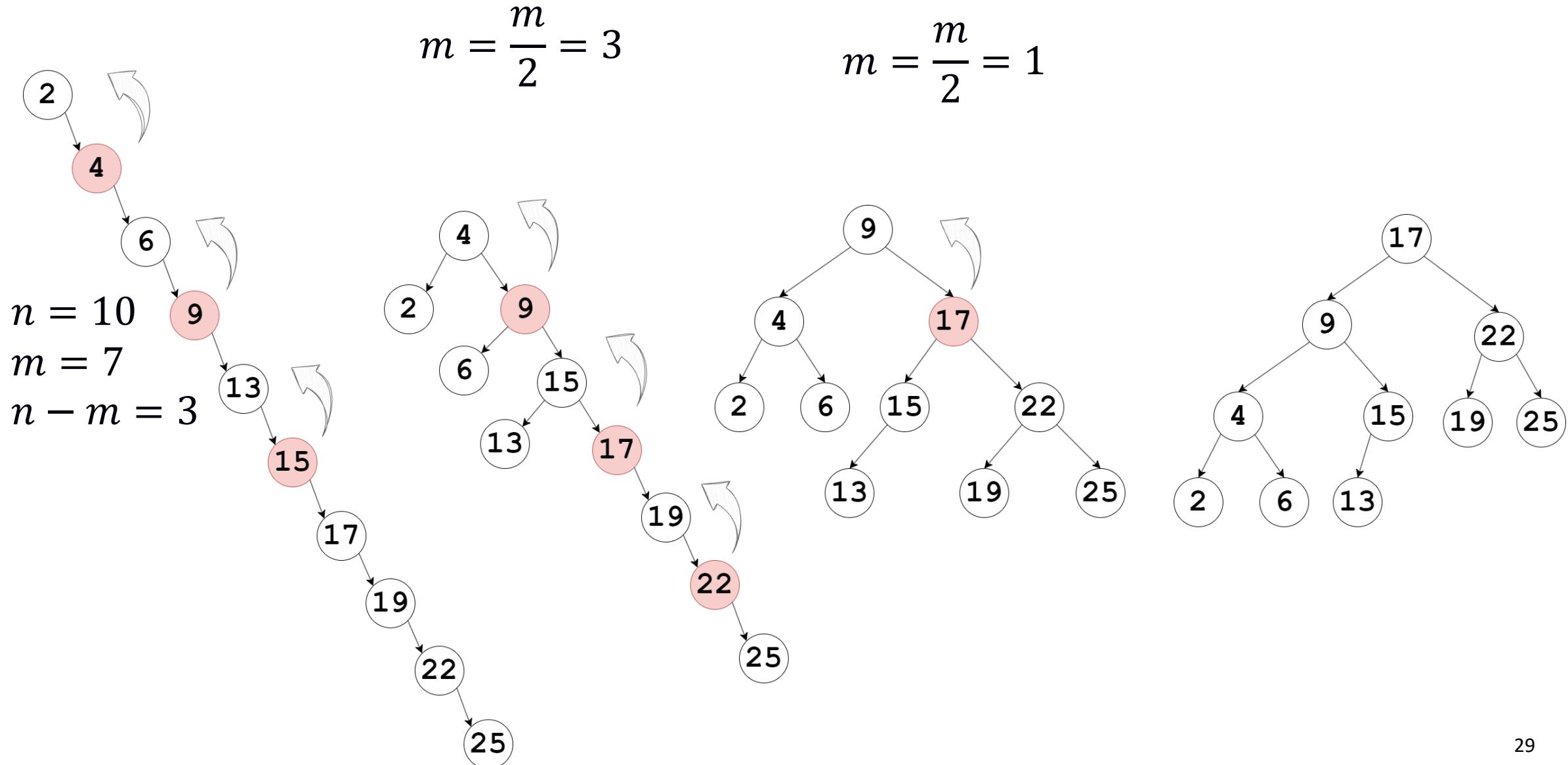
$$m = \frac{m}{2} = 3$$



$$m = \frac{m}{2} = 1$$



# DSW – backbone breaking example 2



# Adelson-Velski-Landis binarno stablo (AVL)

- DSW is offline
- AVL is online
- **Online balancing**
  - Continually check for disbalance after operations
  - Fix the disbalance
- Balances with a minimal impact on structure -> local balancing
- Balances by height

# AVL (2)

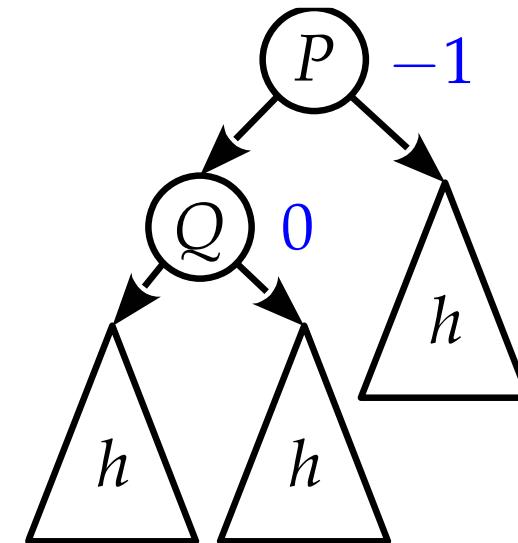
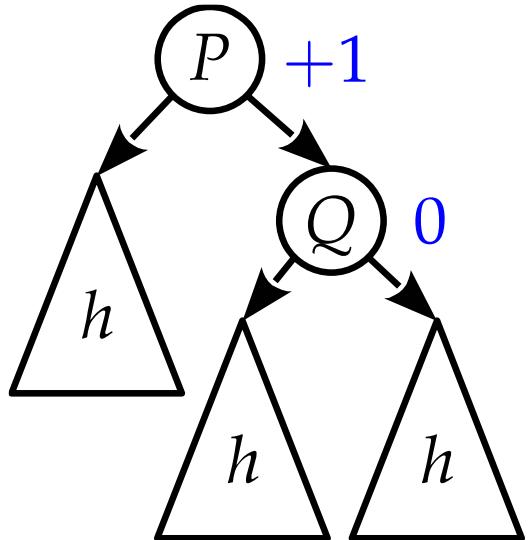
- We check a **balancing factor (BF)** for each subtree

- Recursive

$$BF(S) = h(R) - h(L)$$

- We consider a subtree balanced if:

$$-1 \leq BF(S) \leq 1$$



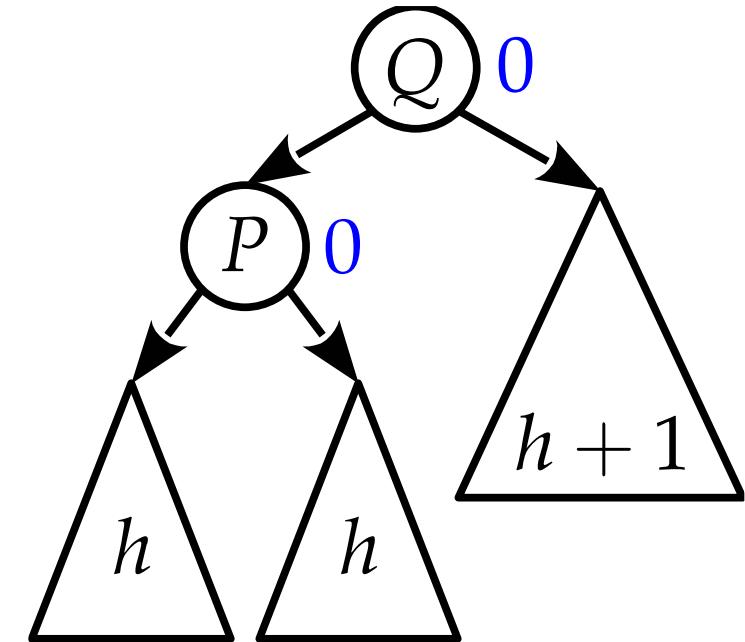
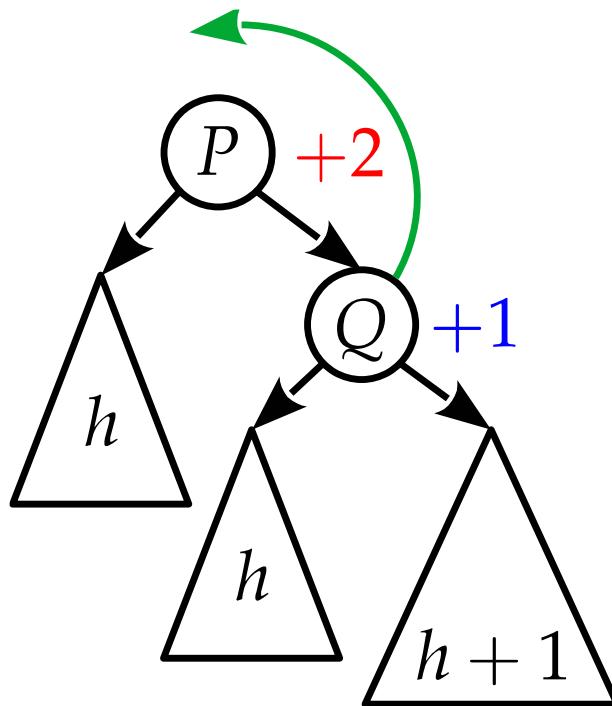
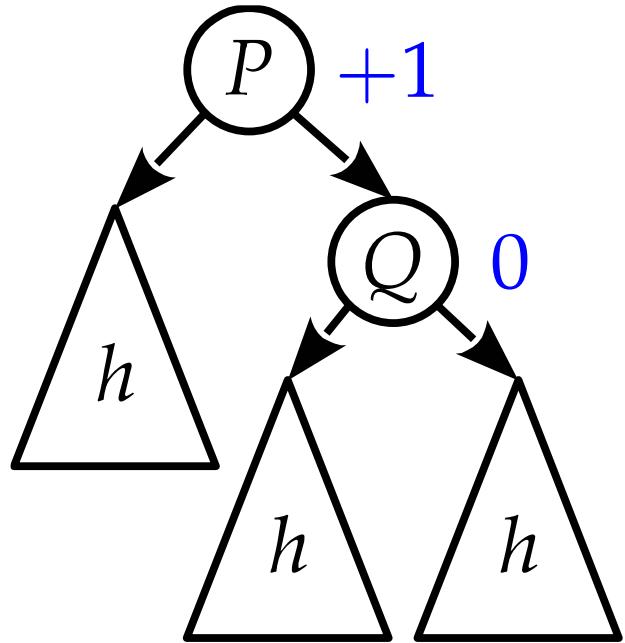
# AVL (1)

- After each operation, we update the BF
- If a  $BF(S) = -2$  or  $BF(S) = 2$  appear, we balance the tree
  - We usually don't let it go far over
- 4 scenarios

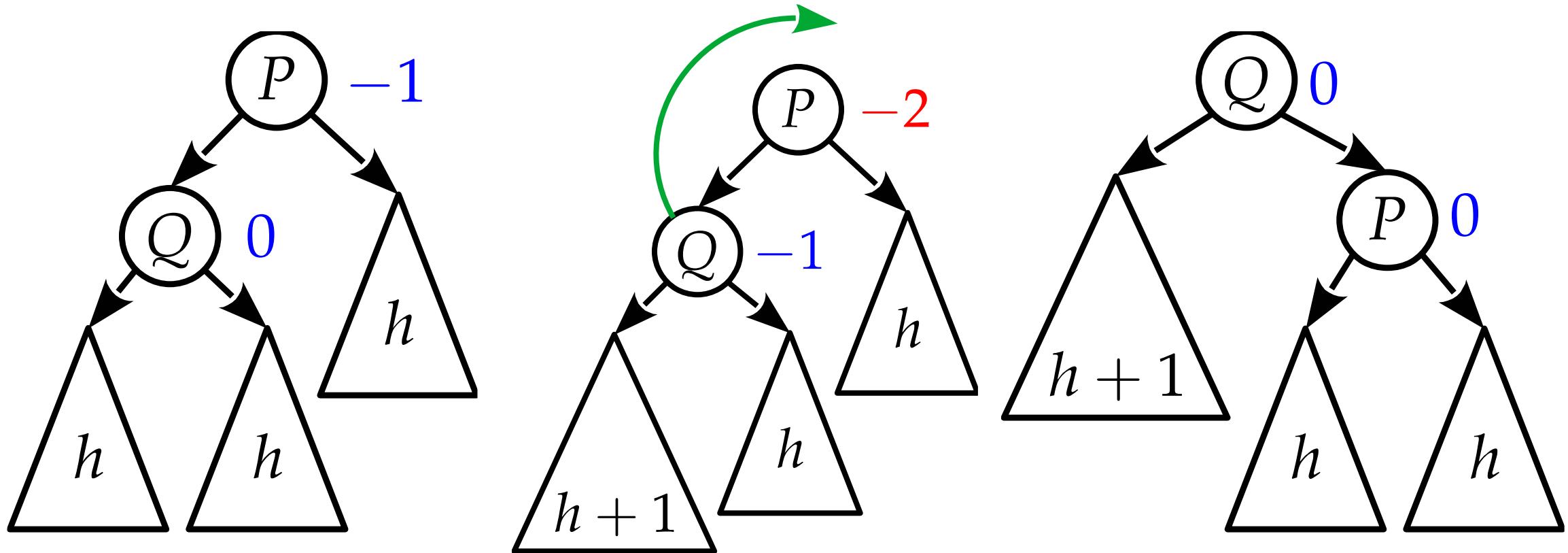
# AVL – Balancing scenarios (2)

Scenario	Condition	Operation
<b>Right Right</b>	higher subtree Z is to the <u>right</u> and has <u><math>BF(Z) \geq 0</math></u>	rotate left X
<b>Left Left</b>	higher subtree Z is to the <u>left</u> and has <u><math>BF(Z) &lt; 0</math></u>	rotate right X
<b>Right Left</b>	higher subtree Z is to the <u>right</u> and has <u><math>BF(Z) &lt; 0</math></u>	rotate right left X
<b>Left Right</b>	higher subtree Z is to the <u>left</u> and has <u><math>BF(Z) \geq 0</math></u>	rotate left right X

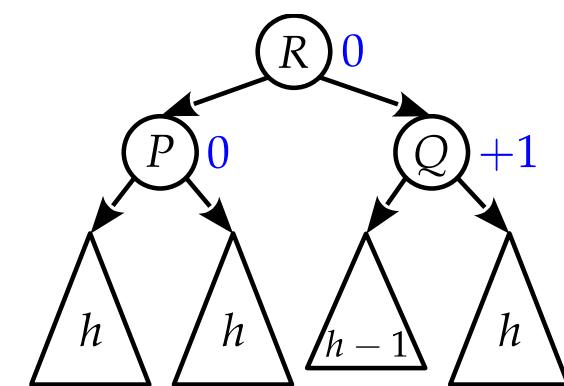
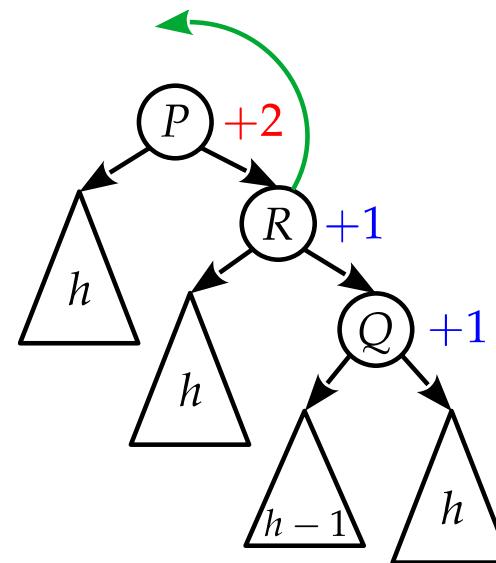
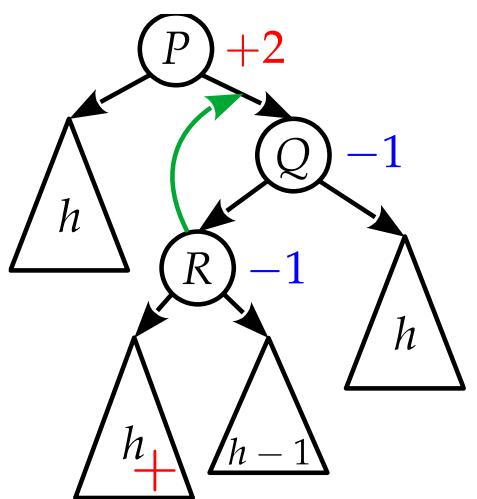
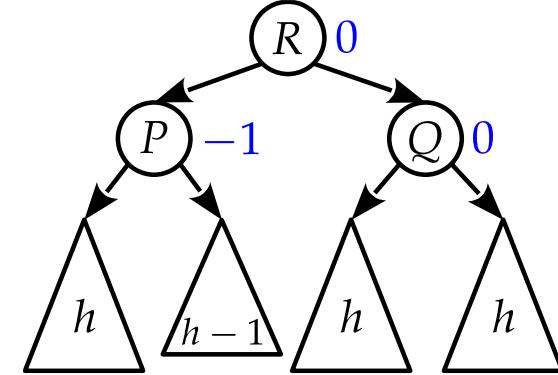
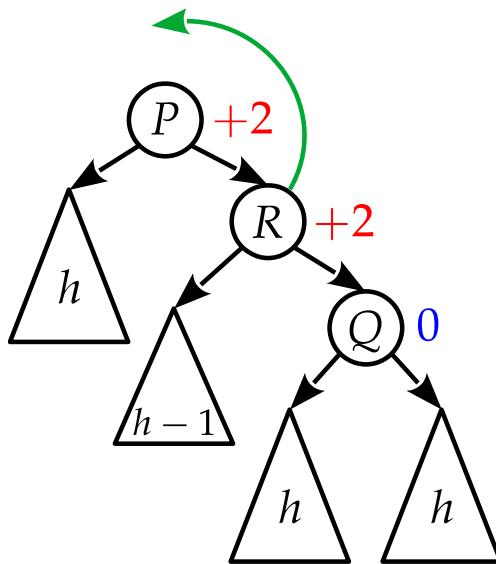
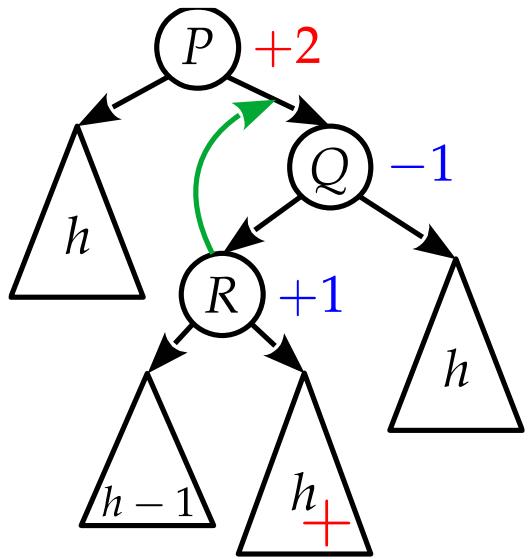
# AVL – Right Right (3)



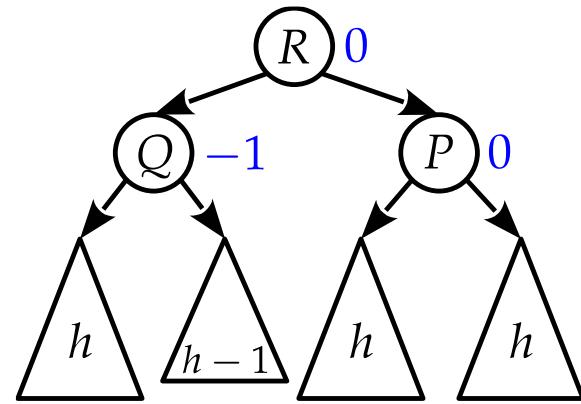
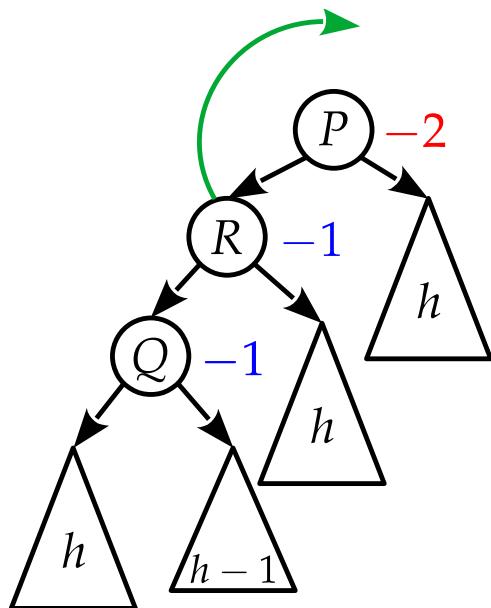
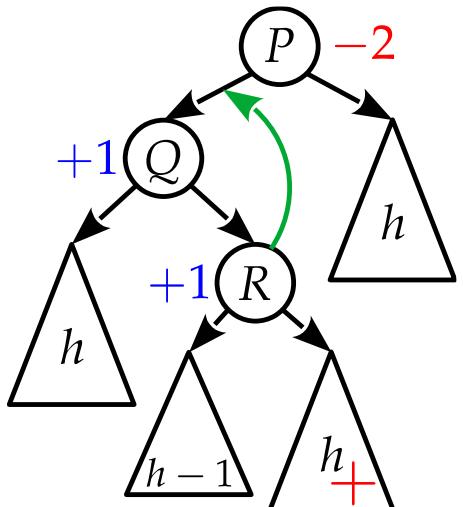
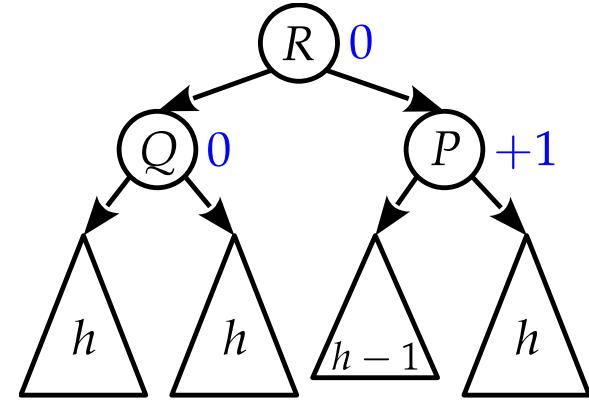
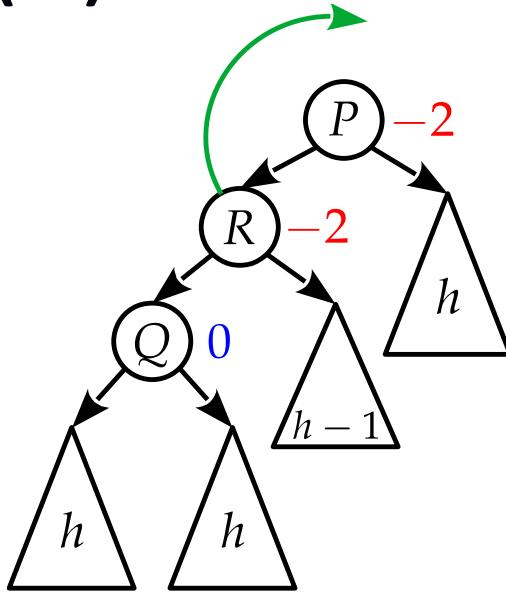
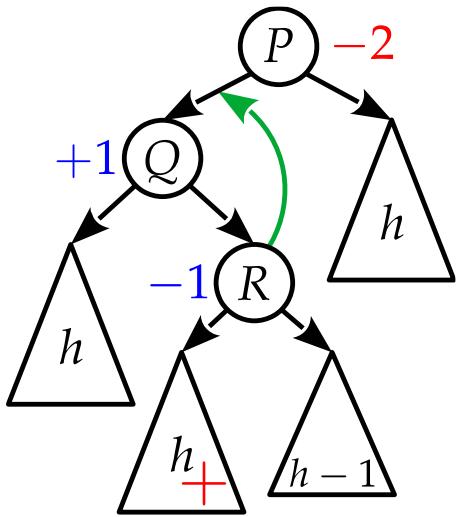
# AVL – Left Left (4)



# AVL – Right Left (6)



# AVL – Left Right (6)



# AVL (7)

- Always copy to delete
  - Least structural change
- Search:  $O(\log_2 n)$
- Insert or delete:  $O(2\log_2 n)$
- Theoretical height of an AVL tree:

$$\log_2(n + 1) \leq h \leq \mathbf{1.44} \log_2(n + 2) - 0.328$$

<https://dl.acm.org/doi/10.5555/525639>

# 02 – Trees 2

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# Red-Black Tree (1)

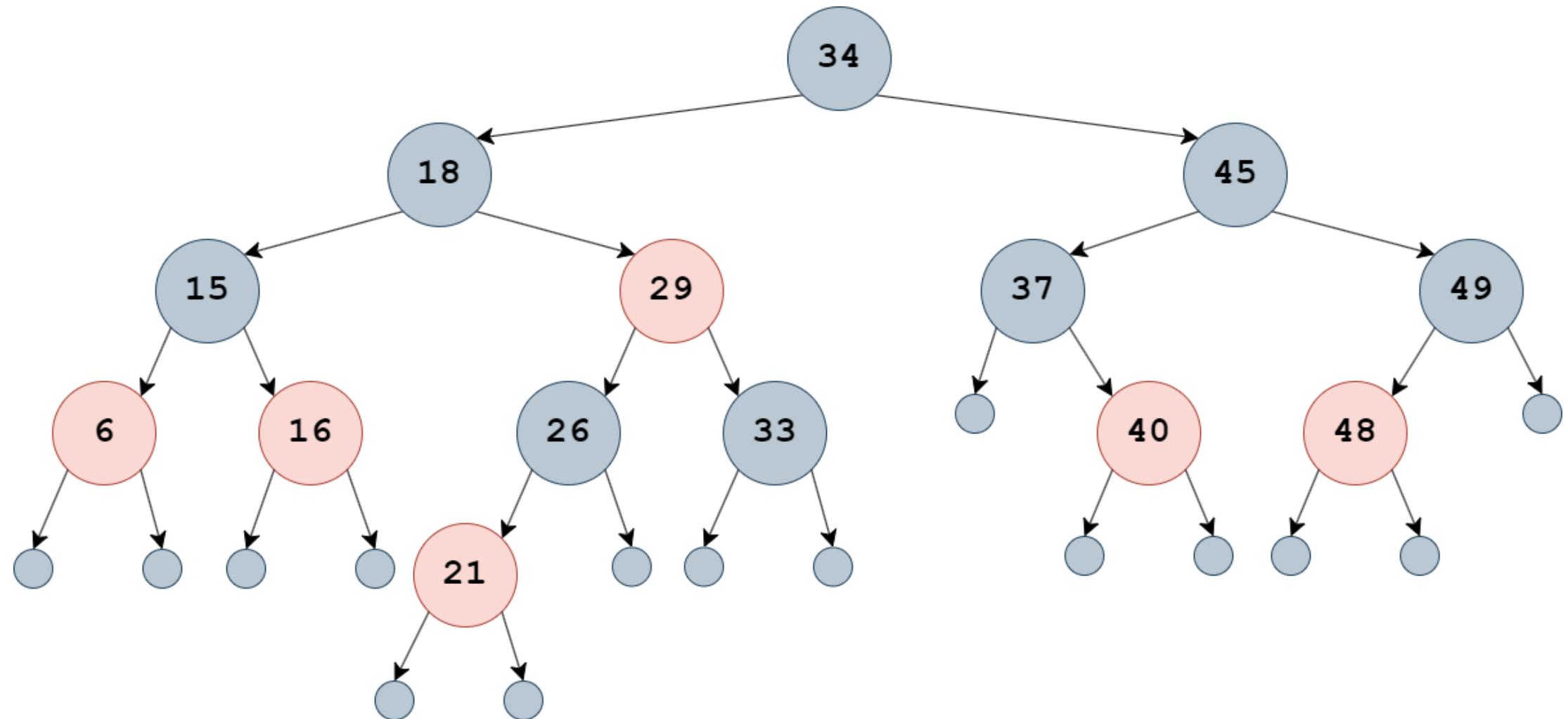
- Binary tree with 2 types of nodes:
  - **RED**
  - **BLACK**
- Constrained by the colors on paths from root to leaf
  - Balanced tree
- Balances horizontally
- Metrics
  - (max.) height:  $h \leq 2\log_2(n + 1)$
  - search complexity:  $O(\log_2 n)$

# Red-Black Tree – definition rules

1. Every node is **red** or **black**
2. Root is **black** (not mandatory; usually switched from red to black)
3. Every leaf – **NIL** – is **black**
4. If a node is **red**, its children are **black**
5. For each node, all simple paths from the node to leaves contain the same number of black nodes.
  - Red and black heights  $h_R(x)$  and  $h_B(x)$  for node  $x$
  - The longest path root->leaf is at most twice as long as the shortest path root->leaf

*Note: NIL leaves don't contain keys so they don't have to \*really\* exist*

# Red-Black Tree – example (3)



# Red-Black Tree - adding a node

1. Add the node like with any sorted binary tree, **but make it Red**
  2. Restructure the tree to adhere to definition rules (DRs)
    - Rotations + Recoloring
    - Adding a node doesn't break DRs 1, 3, and 5
      1. DR 2 is broken if the new node is a root
      2. DR 4 is broken if the parent of a new node is **Red**
- } **Restructuring!**

# RB Tree – restructuring

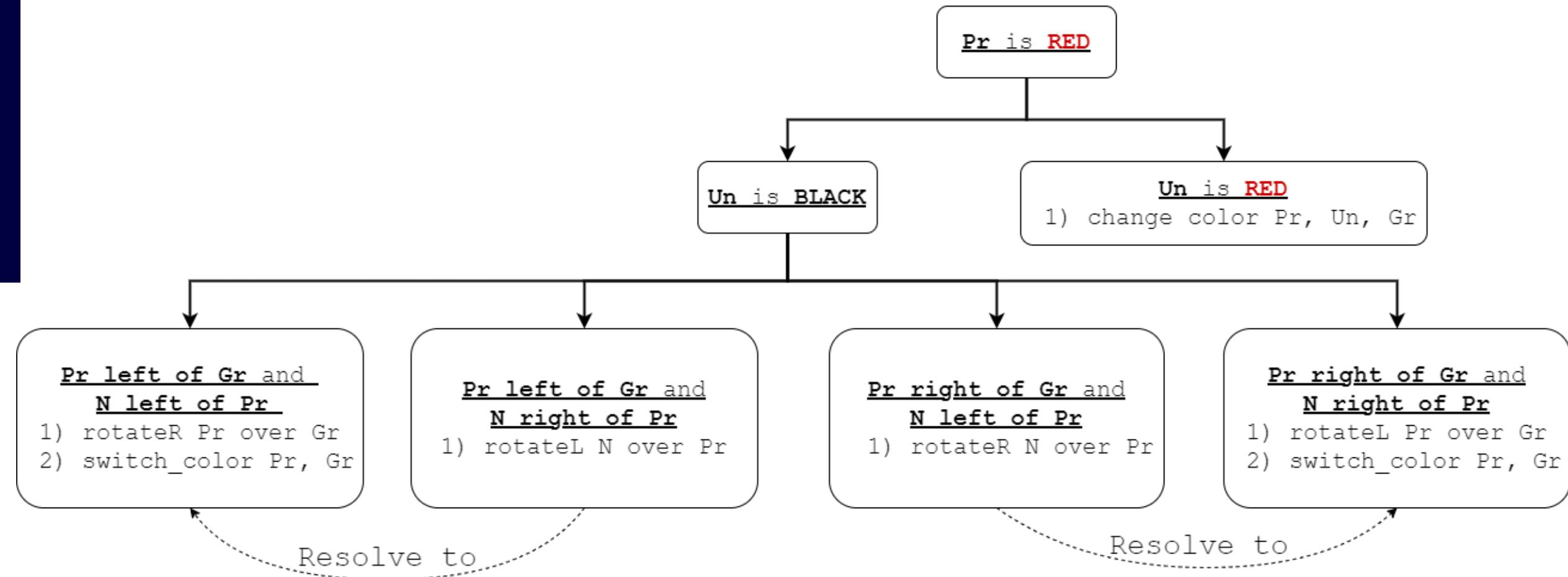
Sc. 1 New node is a root

- DR 2 is broken
- Change the color to **Black**

Sc. 2 Parent of the new node is **Red**

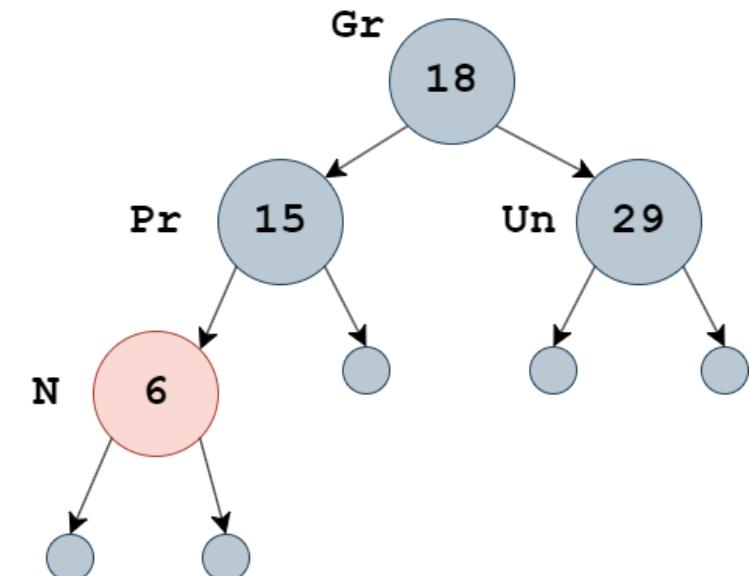
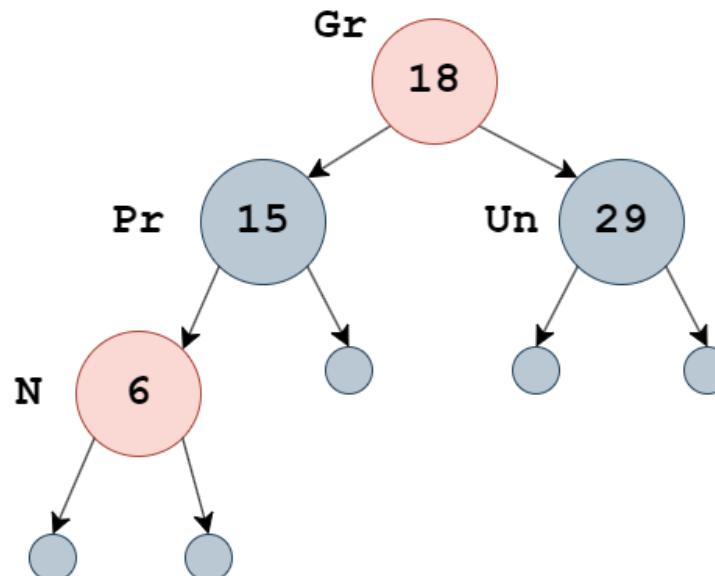
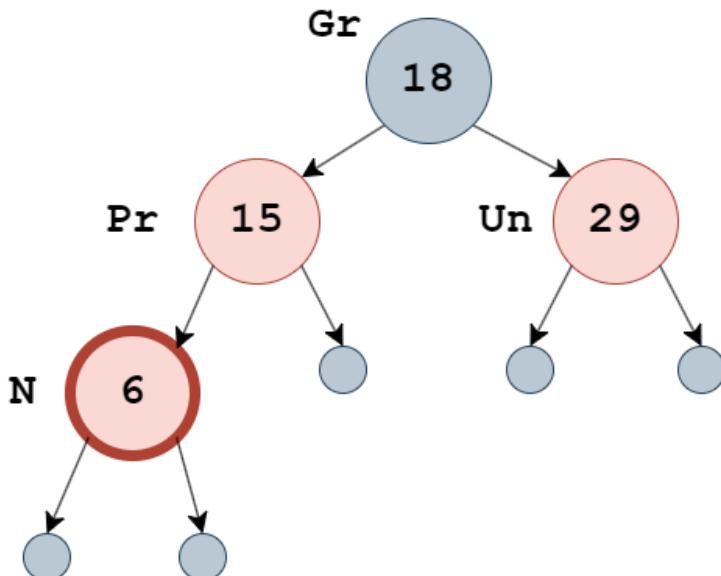
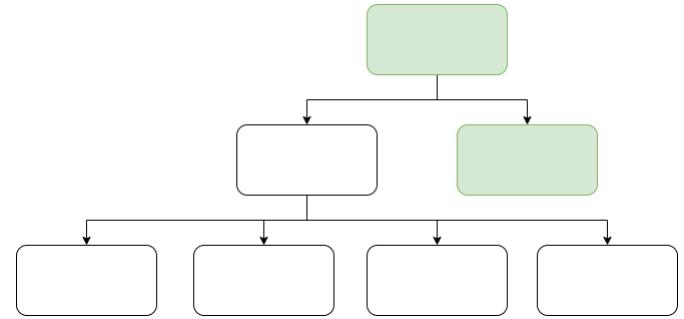
- Decision tree!

# RB Tree – Sc. 2 restructuring



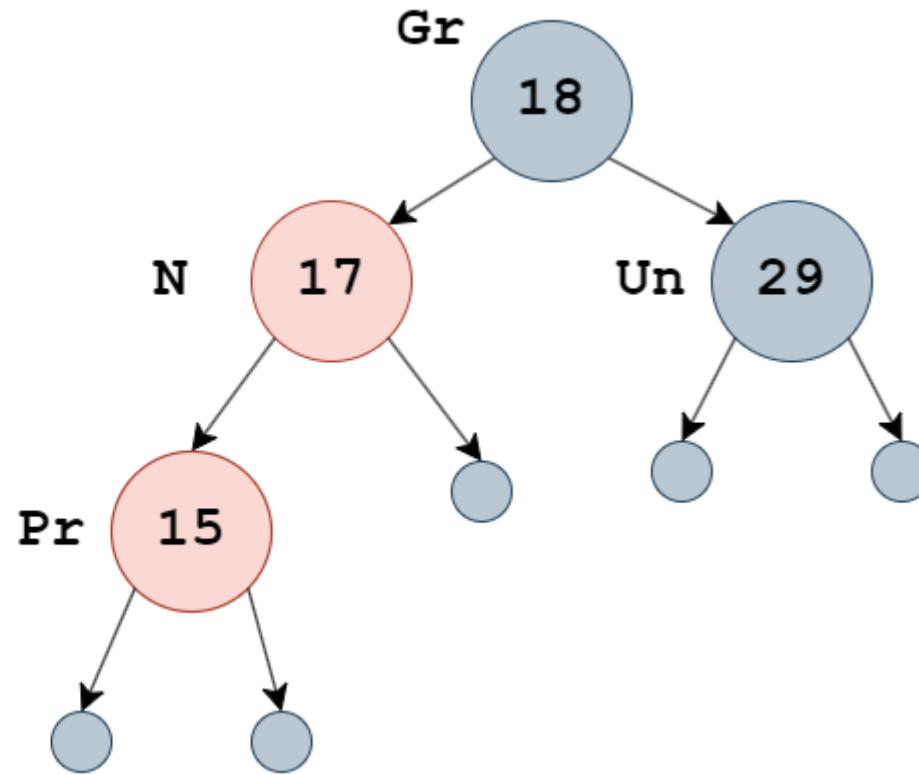
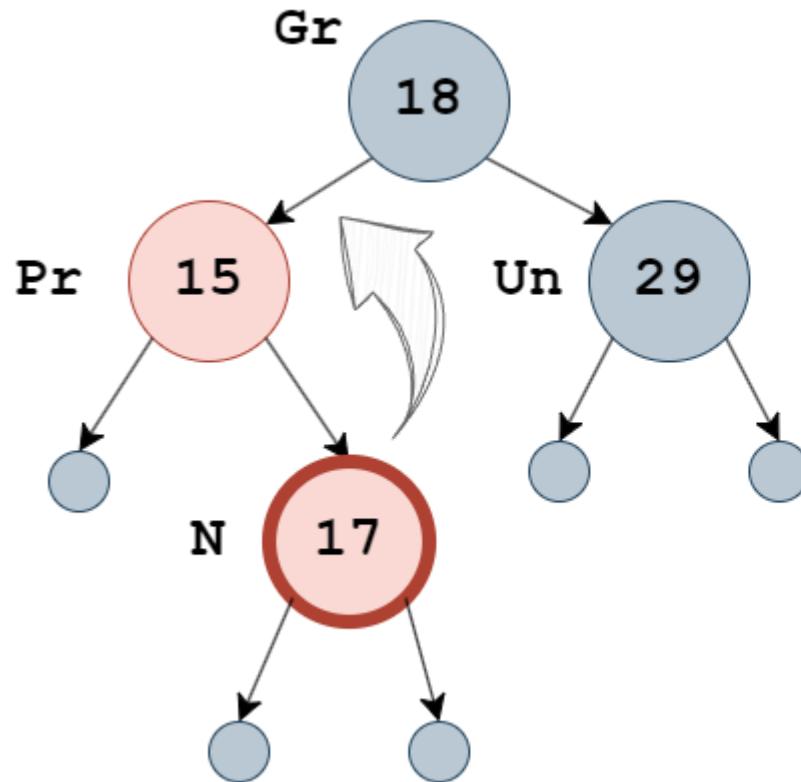
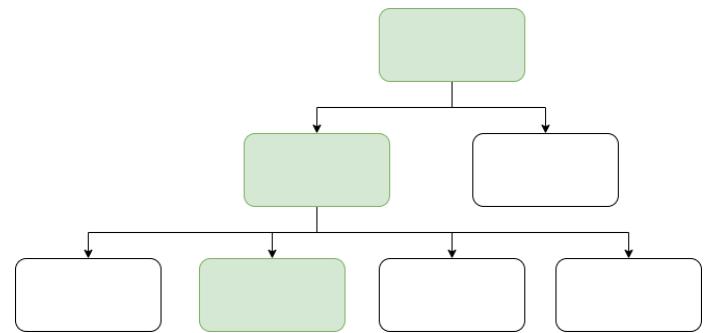
# RB Tree – restructuring (1)

1. Added 6 as **red**
2. Switch colors of Pr, Un, and Gr
3. Switch back root to be **black**

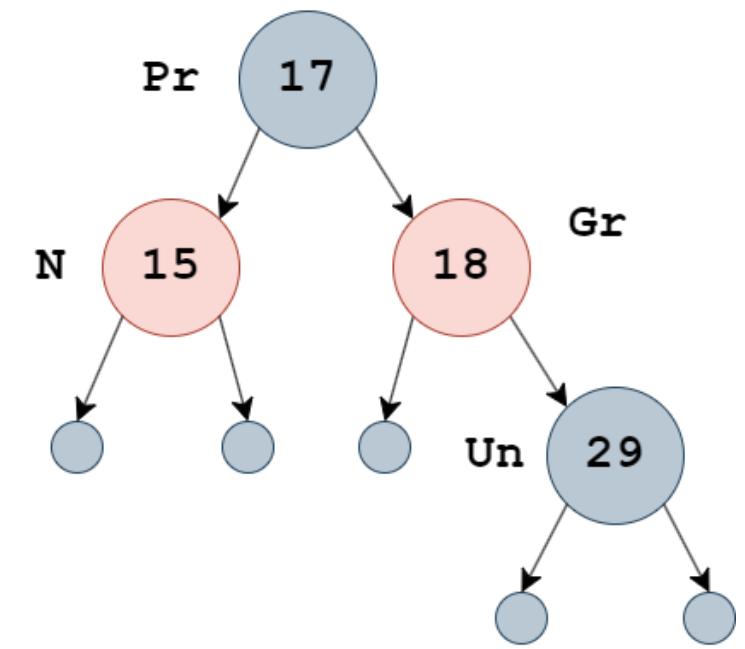
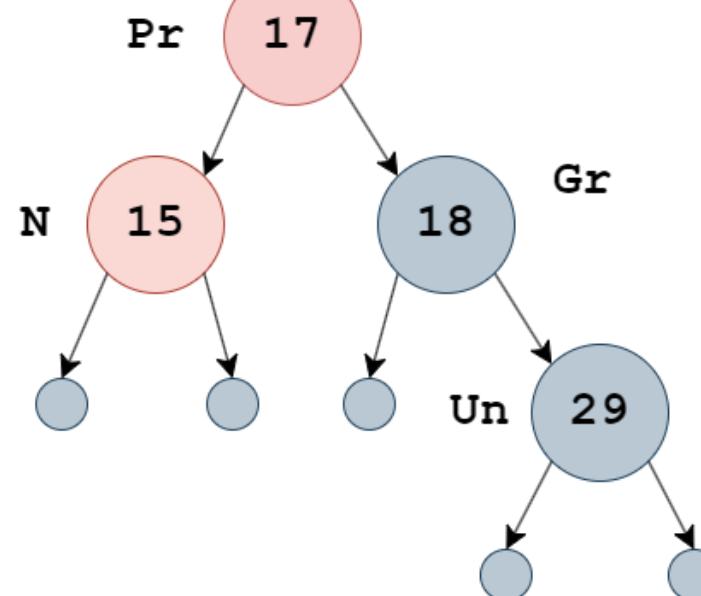
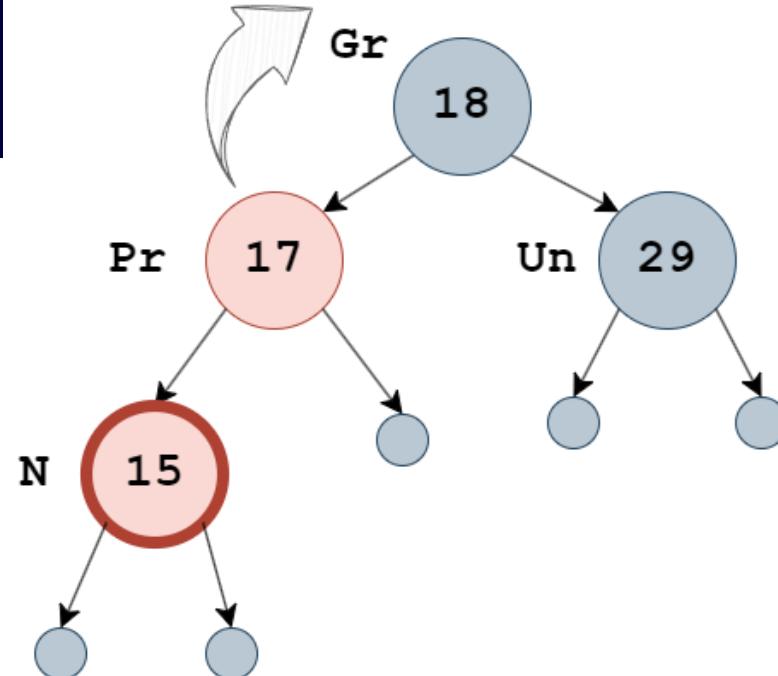
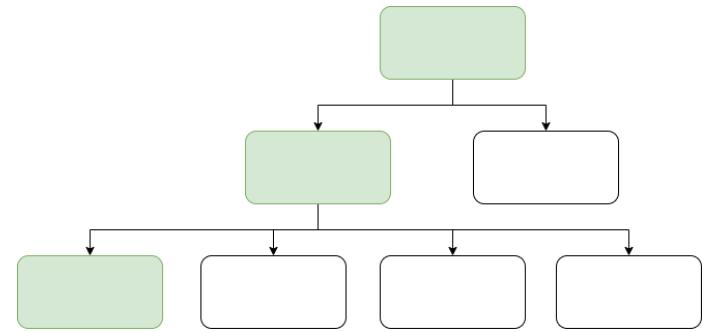


# RB Tree – restructuring (2)

New case!



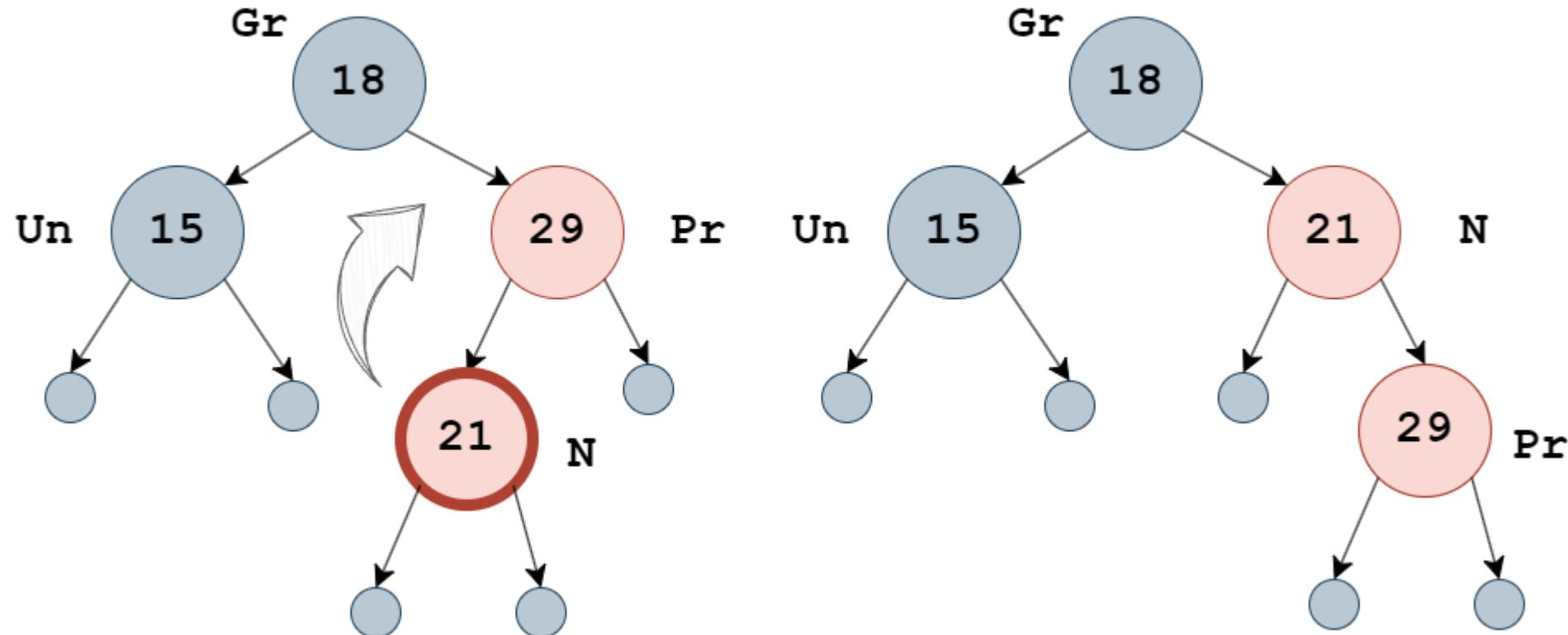
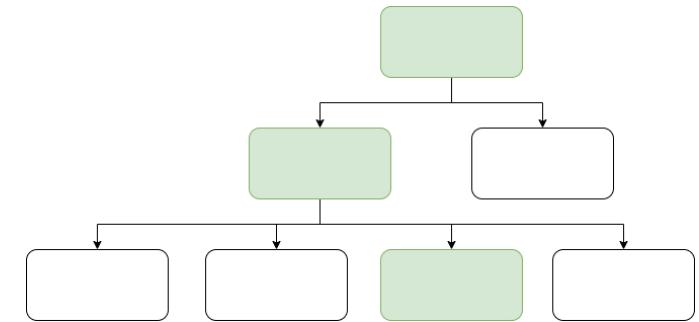
# RB Tree – restructuring cont. (3)



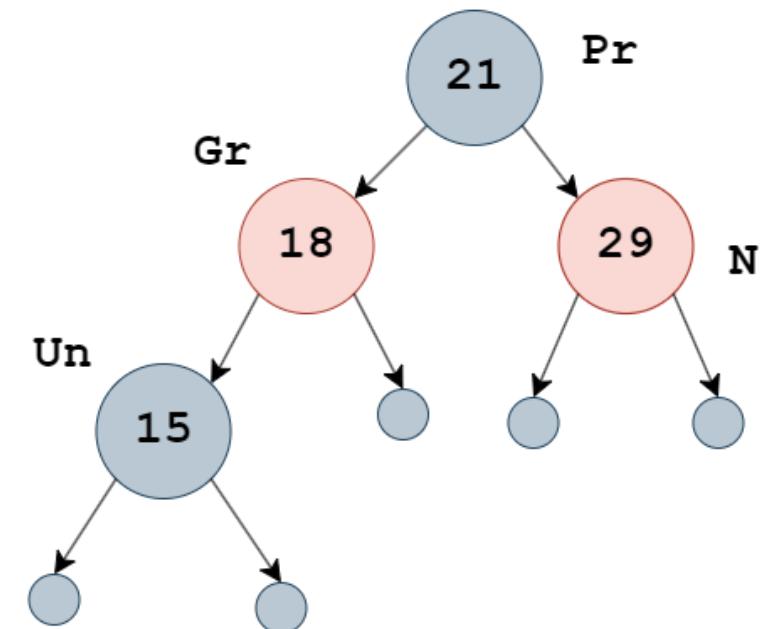
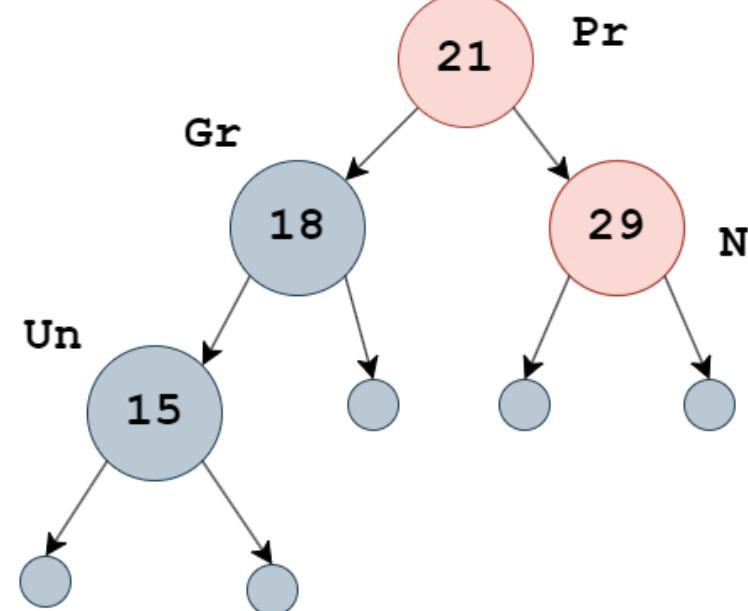
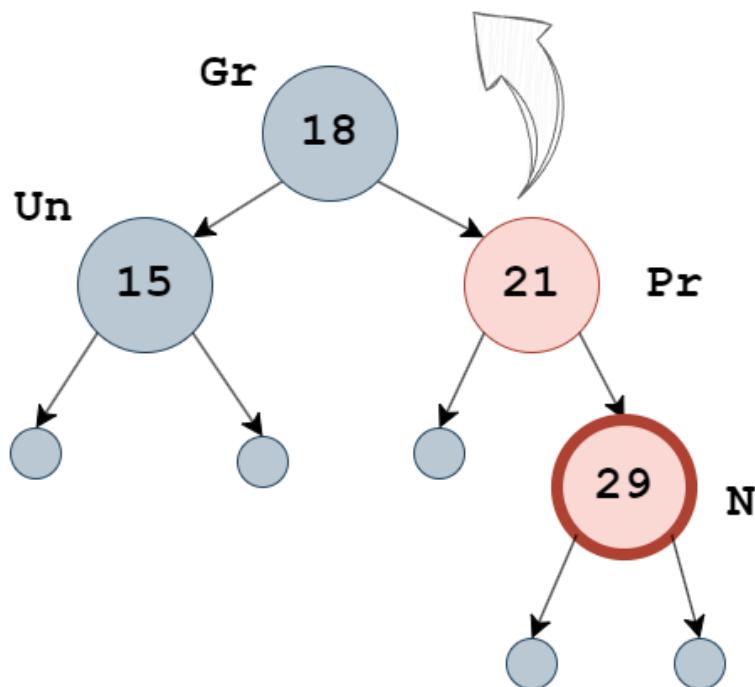
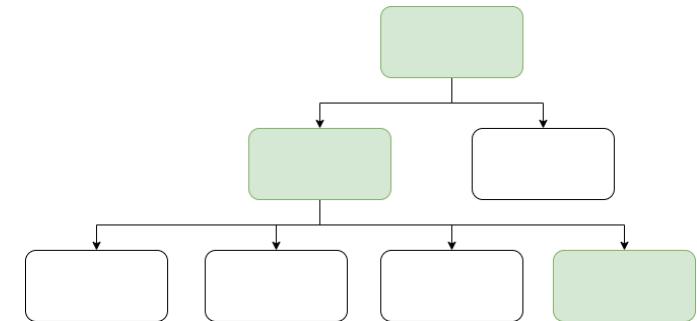
\*Note change in relation names

# RB Tree – restructuring (4)

Symmetric case!



# RB Tree – restructuring cont. (5)



\*Note change in relation names

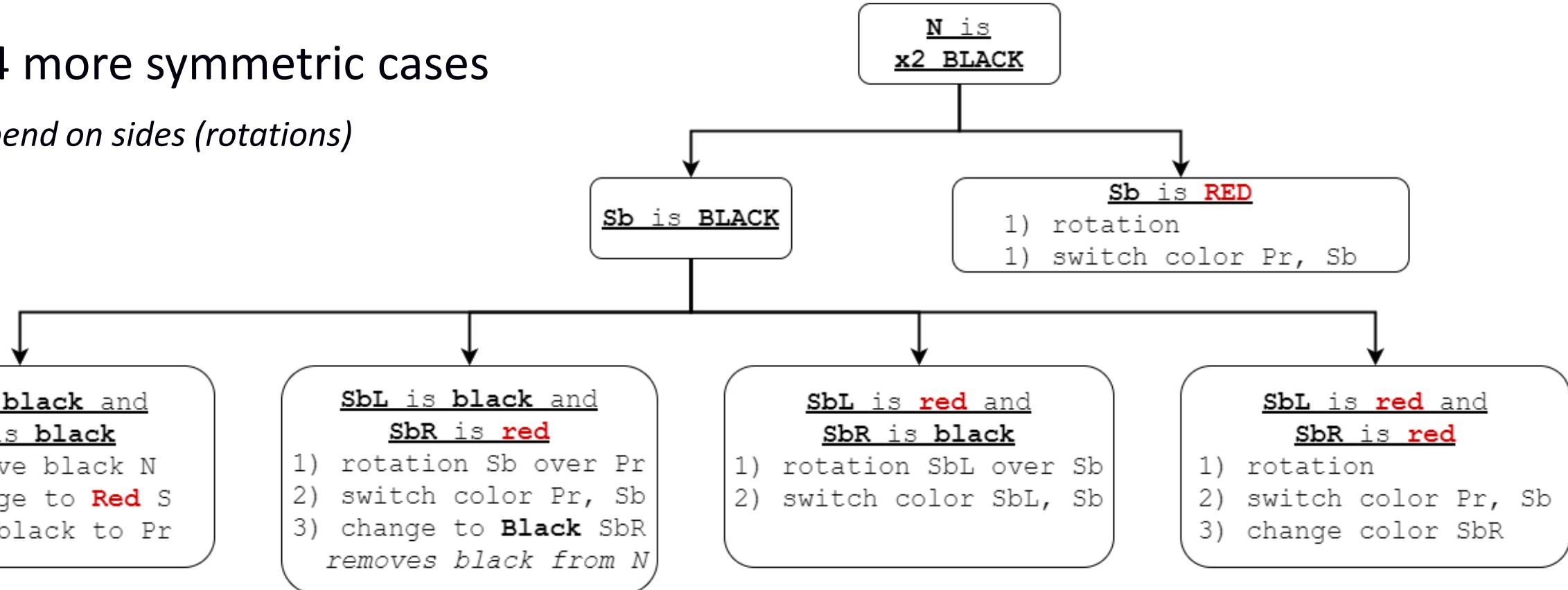
# RB Tree – deleting a node (1)

1. Delete by copying
  2. Remove replacement node
    - If **Red** – done!
    - If **Black** – a complex procedure...
- 
- **Double black** marking
    - Root node get **x2 black** removed automatically

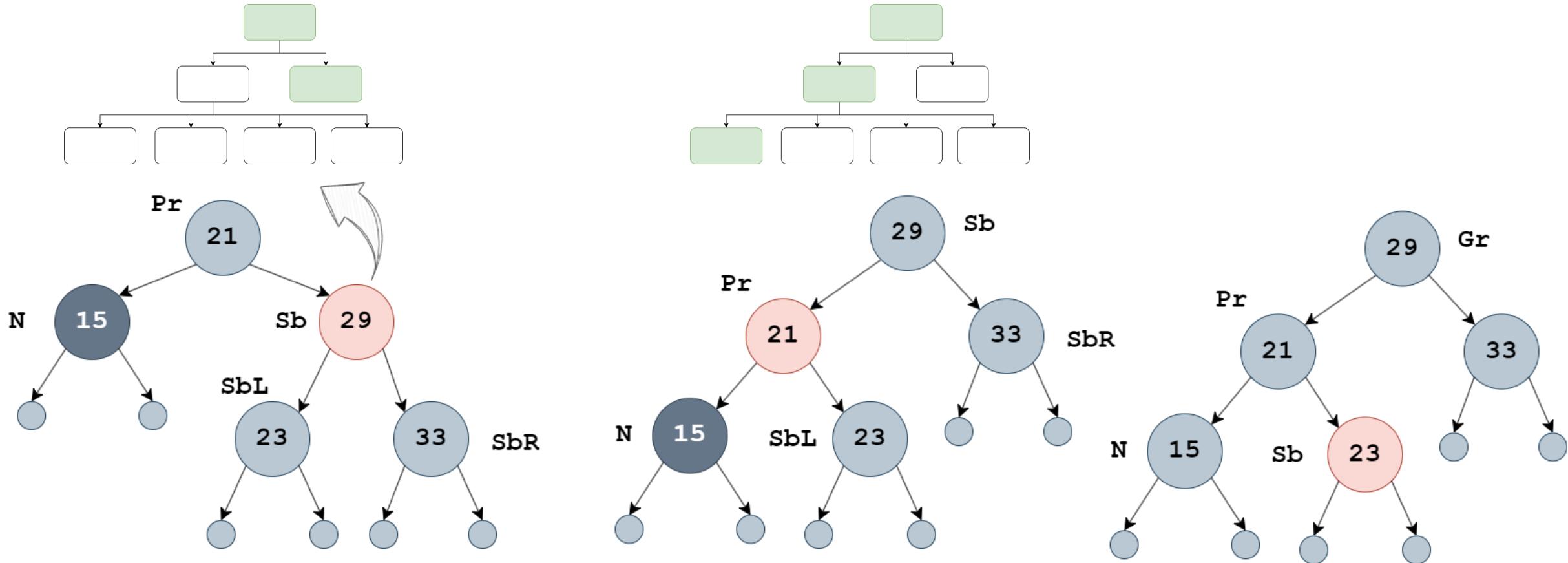
# RB Tree – deleting a node (2)

+ 4 more symmetric cases

*depend on sides (rotations)*



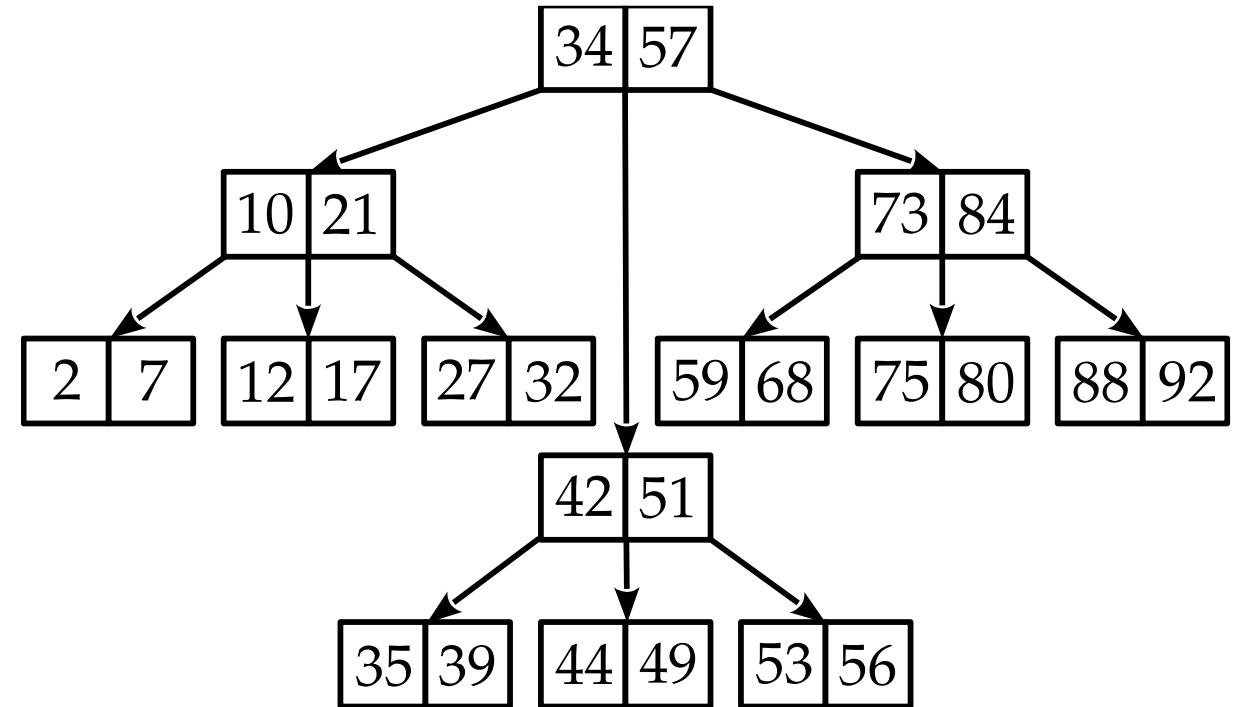
# RB Tree – deleting a node (3)



*Note: renaming after step 2*

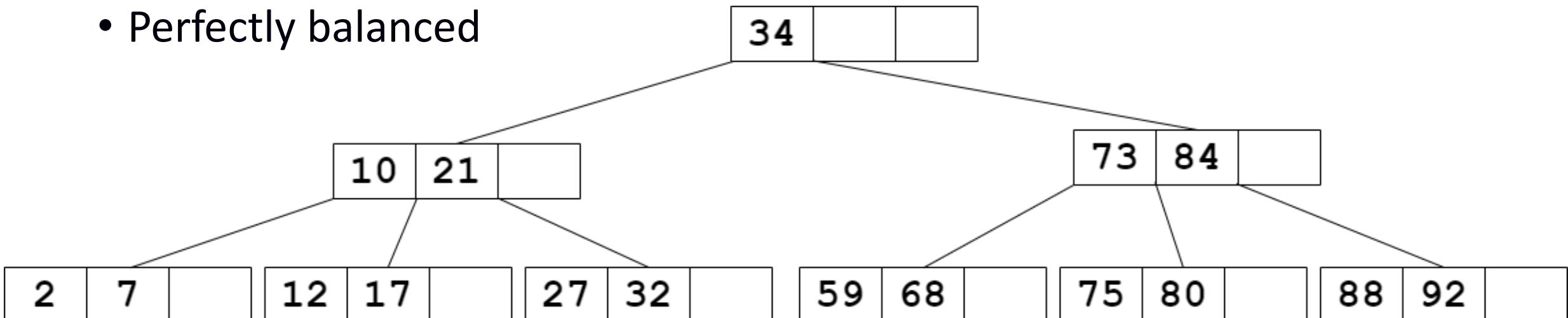
# Multiway tree

- Mth order of children
- M children
- M-1 data



# B-tree (1)

- Special case of M-tree
- Good for sequential memory read
  - Database indexes
- At least 50% capacity filled – not including the *root*
- Perfectly balanced



# B-tree (2)

- M-tree with added features:

1. Root has at least 2 children; unless root = leaf
2. Every node has  $x-1$  keys and  $x$  sub-tree pointers, while:

$$\lceil m/2 \rceil \leq x \leq m$$

3. All leaves have at least  $x-1$  keys, while:

$$\lceil m/2 \rceil \leq x \leq m$$

4. All leaves are at the same level – perfectly balanced

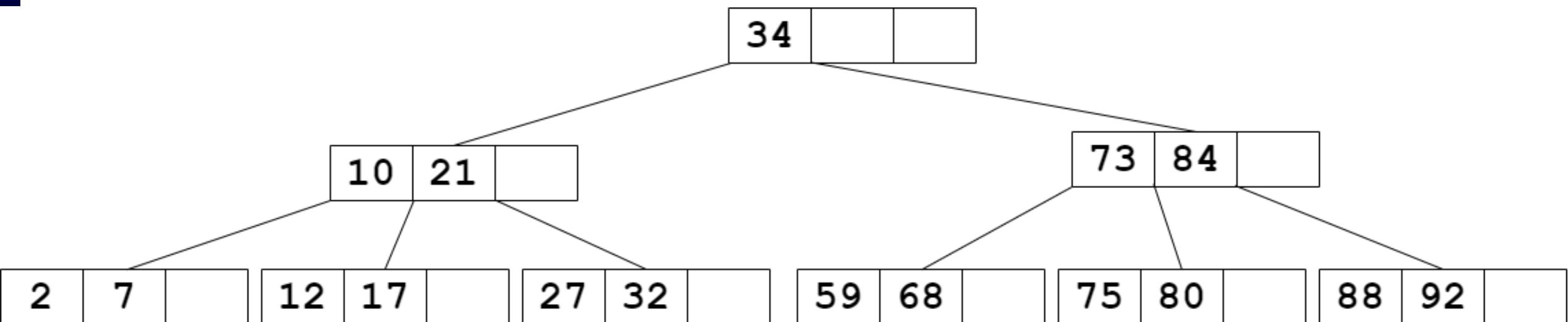
# B-tree – search (1)

Searching for a key for a value

1. Enter a node and seq. read keys while:
  - key > value AND there are unread keys
  - HALT if key == value
2. If value > key OR end of the node is reached
  - Descend to the lower level
  - If there is no lower level – KEY NOT FOUND

# B-tree – search (2)

- Find 17, 59, 84, 18

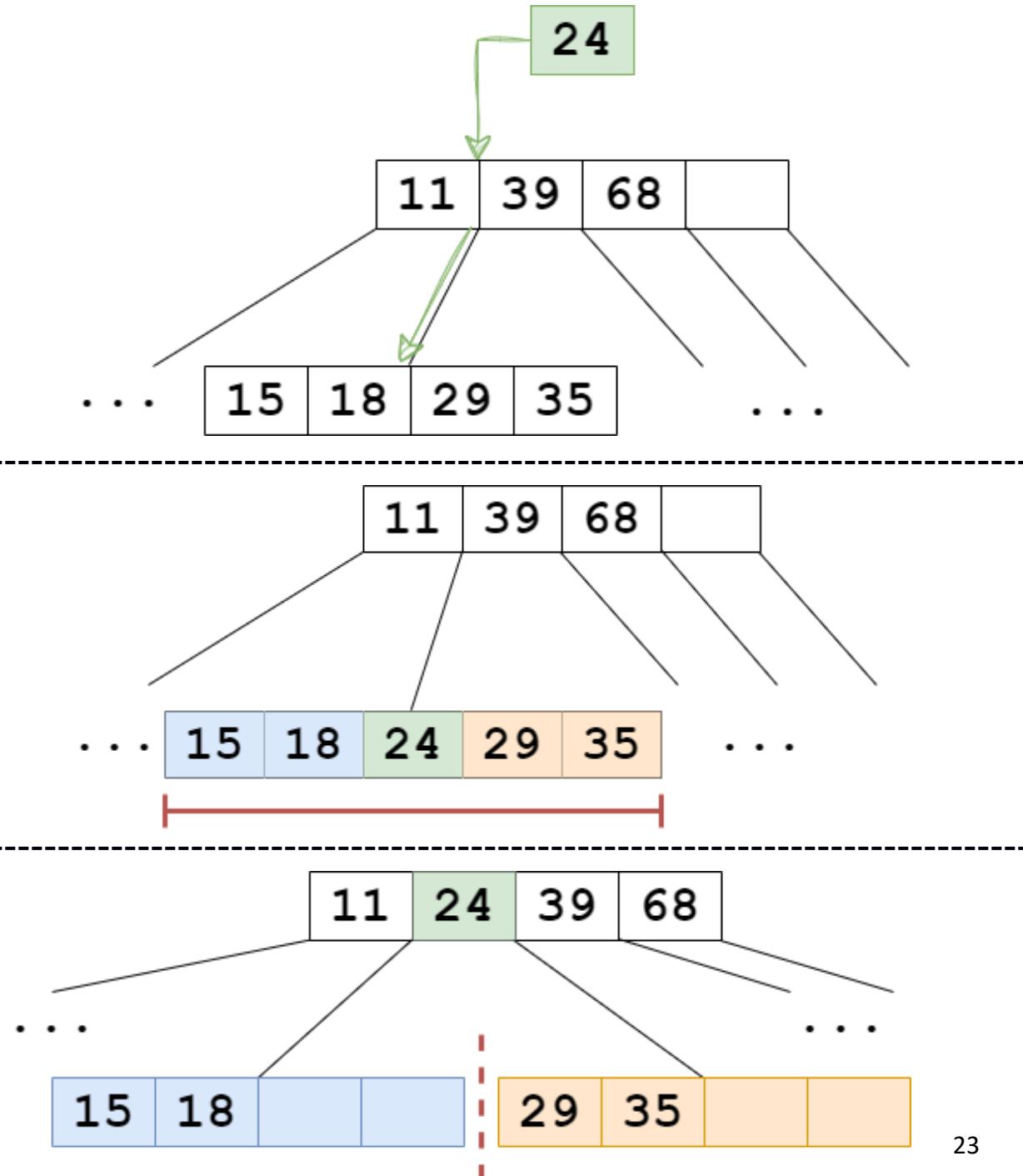


# B-tree – insertion

1. Search for the appropriate leaf node
2. If there is space in the leaf, place the key
3. If there is no space in the leaf, split the node equally into two nodes
  - The middle key goes into the parent node of the leaf node
4. If the parent node has no space, continue the splitting upward

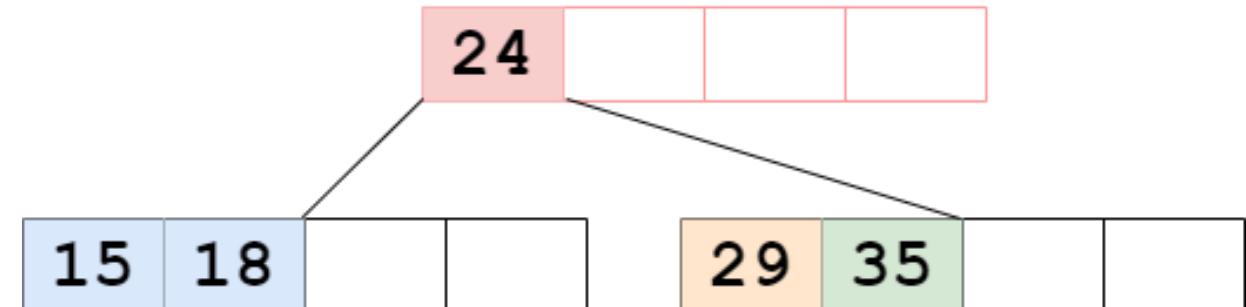
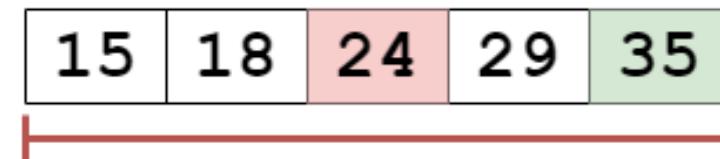
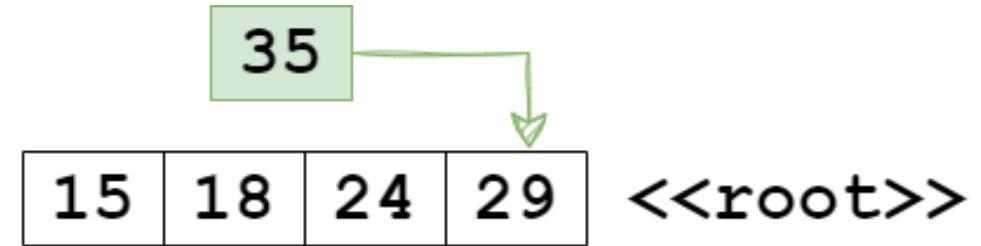
# B-tree – splitting while inserting (1)

- $m = 5 \rightarrow m-1$  keys and  $k$  pointers
- Add 24
- $n > m-1 ?$ 
  - Split!



# B-tree – splitting while inserting (2)

- Splitting a root node
- Preemptive splitting



# B-tree – deletion (1)

**Sc. 1** – target is a leaf key

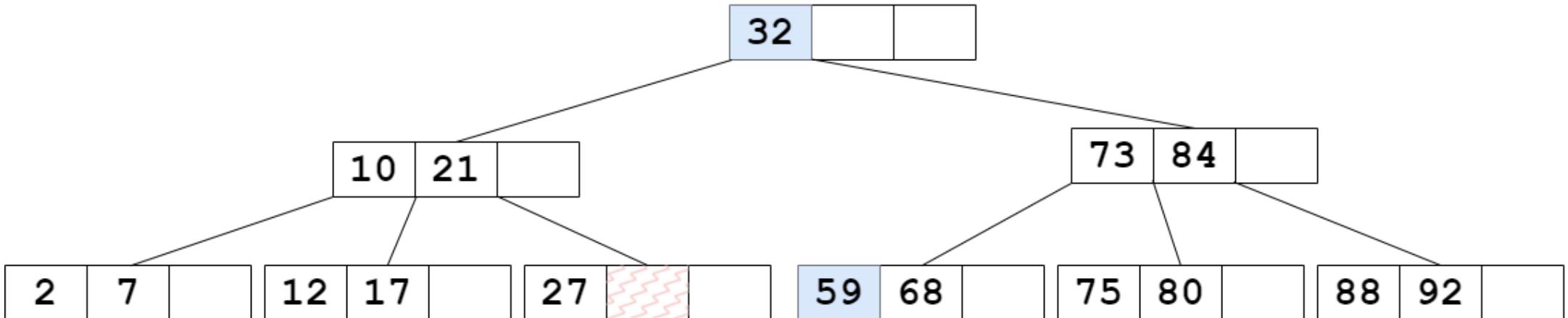
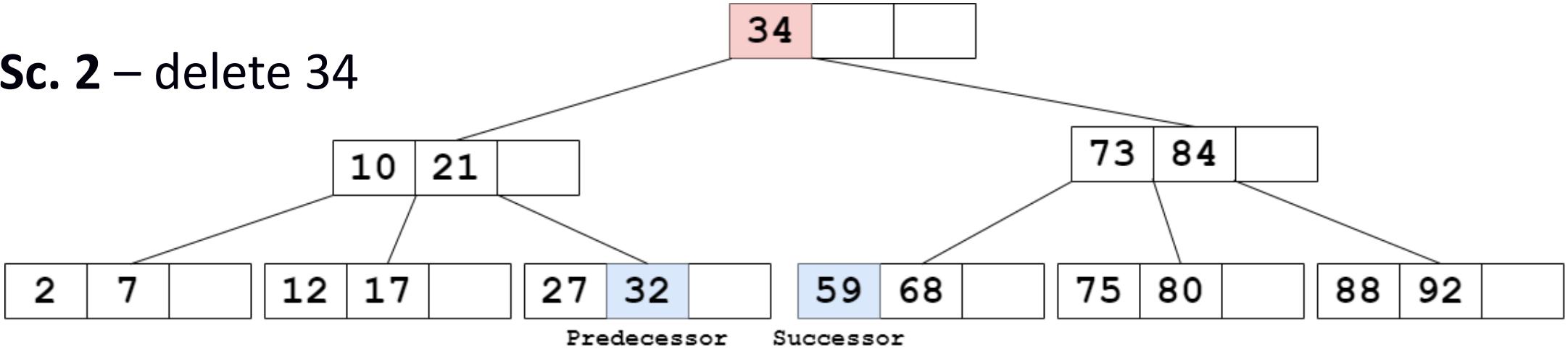
- Delete leaf key

**Sc. 2** – target is a root key

- Find a **predecessor** (closest value in a leaf)
- Delete the leaf key and place the predecessor key into it (eff. delete by copy)
- Preemptive merging
  - While searching – “steal” greatest key from ChL or least key from ChR

# B-tree – deletion (2)

Sc. 2 – delete 34

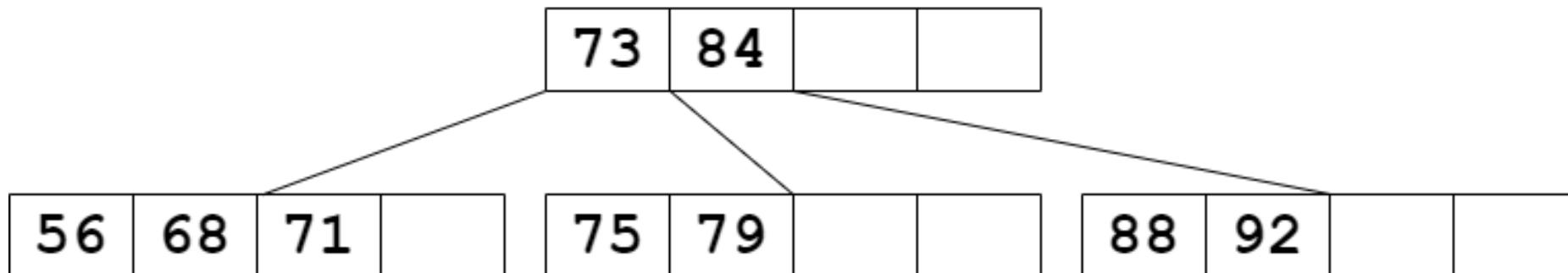
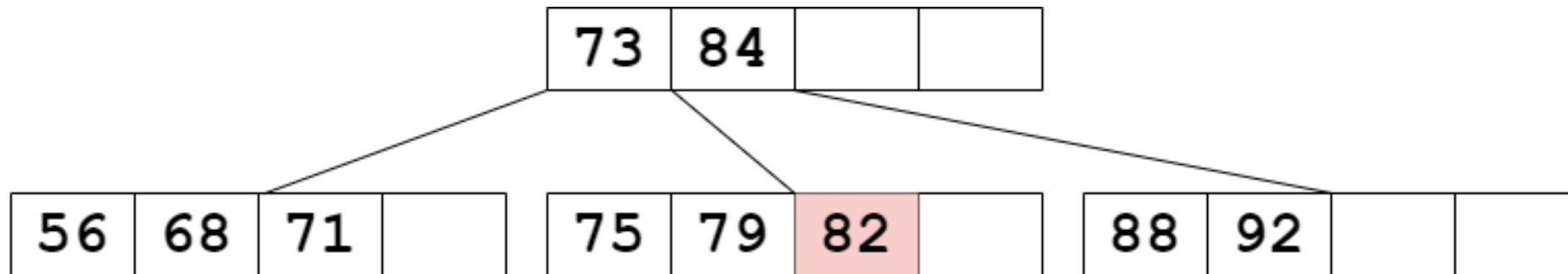


# B-tree – state after deletion

1. Leaf still has  $N$  keys  $\geq \lceil m/2 \rceil - 1 \rightarrow \text{HALT}$
2. Leaf has  $N$  keys  $< \lceil m/2 \rceil - 1$ 
  - **IF** SbL or SbR with  $N$  keys  $> \lceil m/2 \rceil - 1$ 
    - Union of target leaf keys, Pr key, and Sb keys(order is important)
    - Separate into new equally filled SbL' and SbR' and a (new) Pr key  $\rightarrow \text{HALT}$
  - **ELSE** Union of target leaf, Sb, and Pr key
    - They form a new leaf  $\rightarrow \text{EVALUATE Pr}$
  - **ELSE IF** target is *root*
    - **IF** root has  $N$  keys  $> 1$ : Union of target and Sb  $\rightarrow \text{HALT}$
    - **ELSE** Union of target, sb, and root into *new root* node  $\rightarrow \text{HALT}$

# B-tree – deletion example (1)

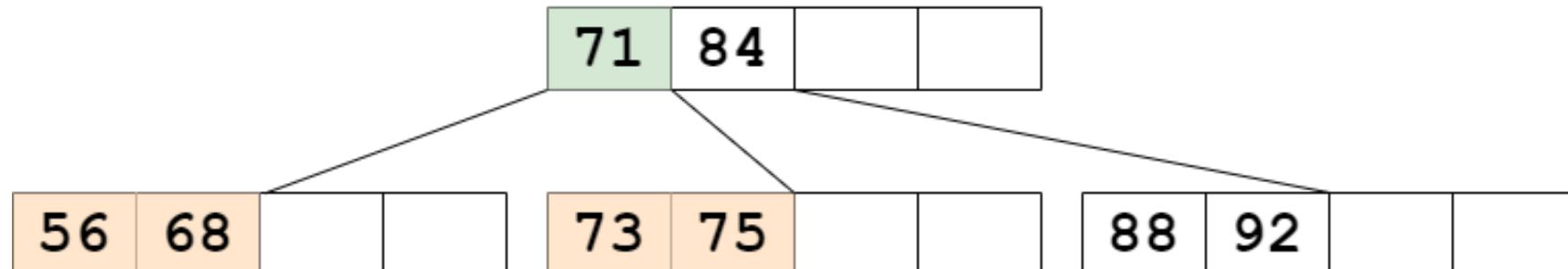
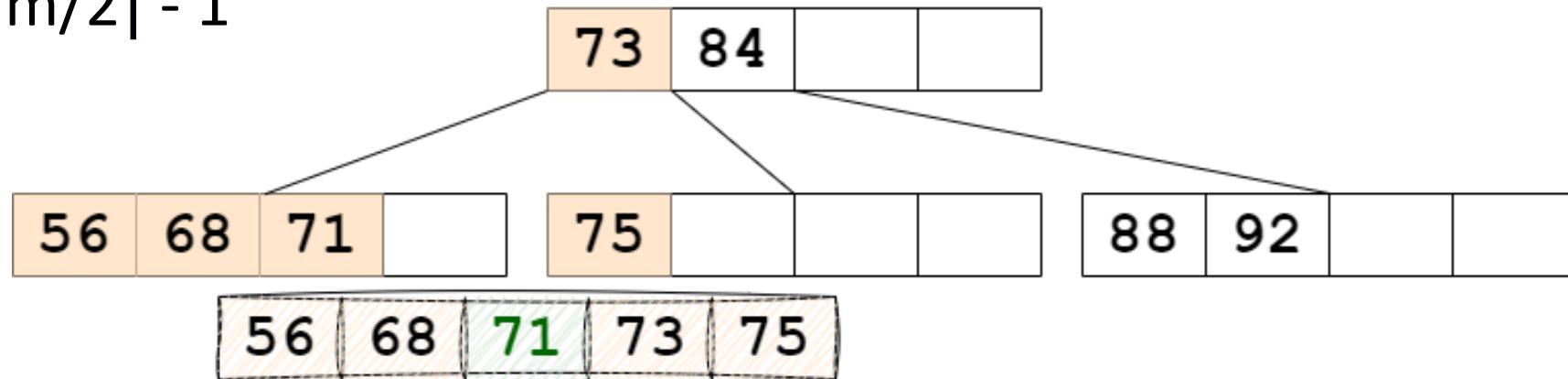
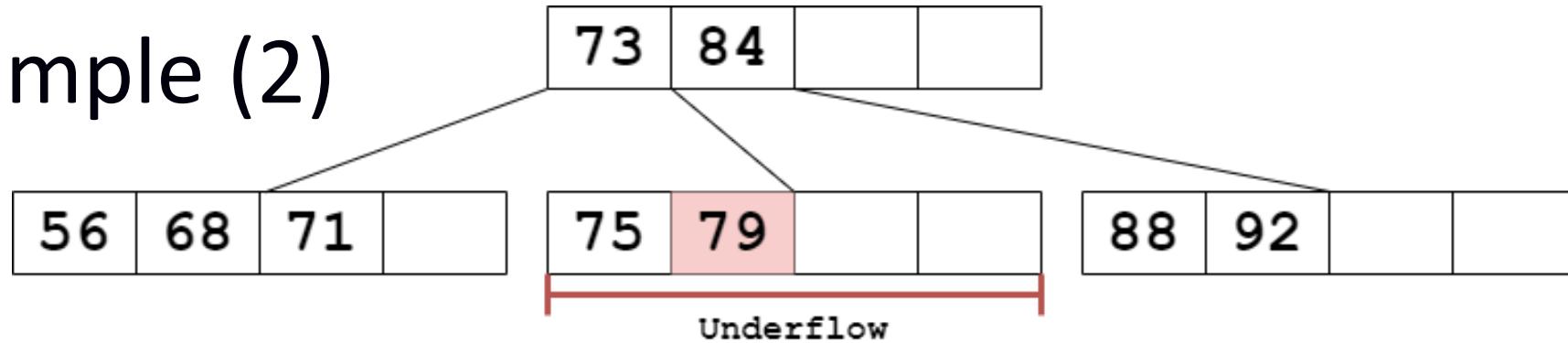
- Delete 82
- Leaf still has  $N$  keys  $\geq [m/2] - 1$



# B-tree

## – deletion example (2)

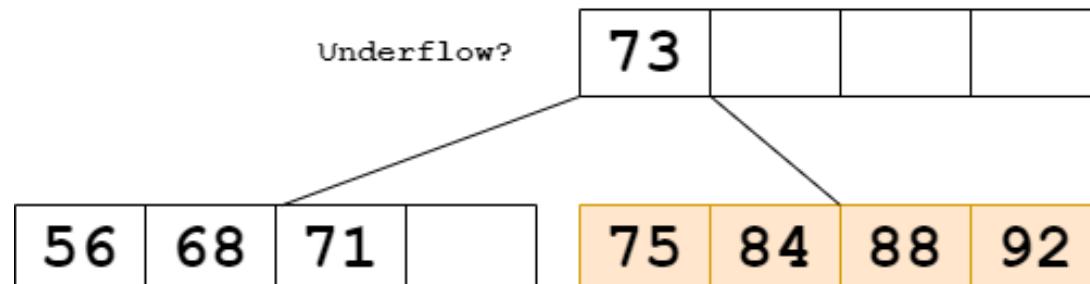
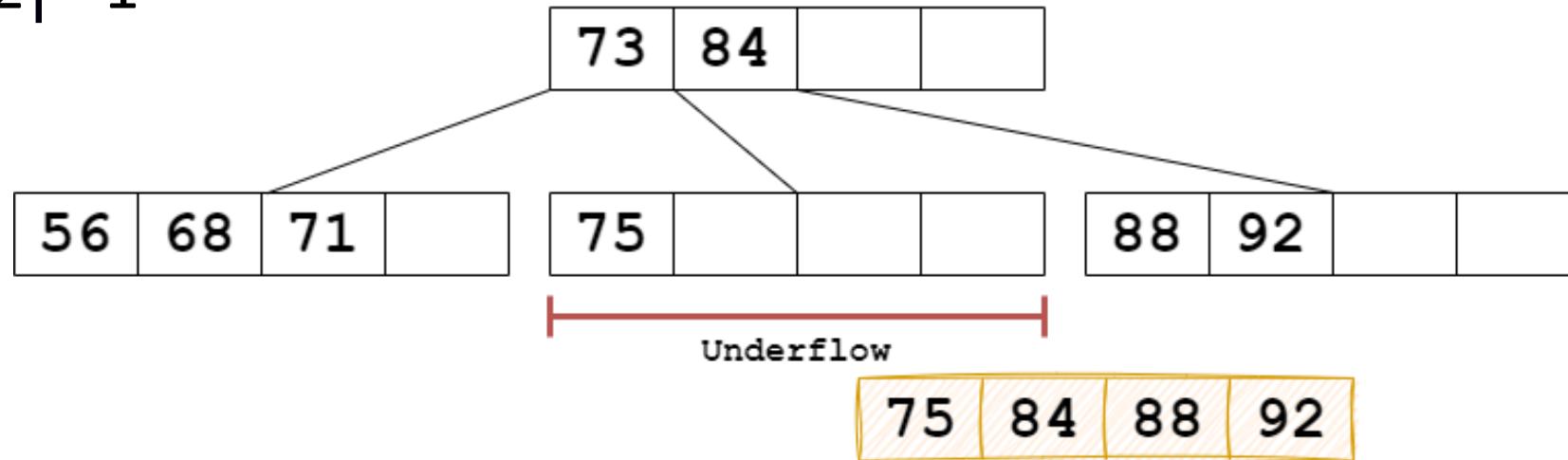
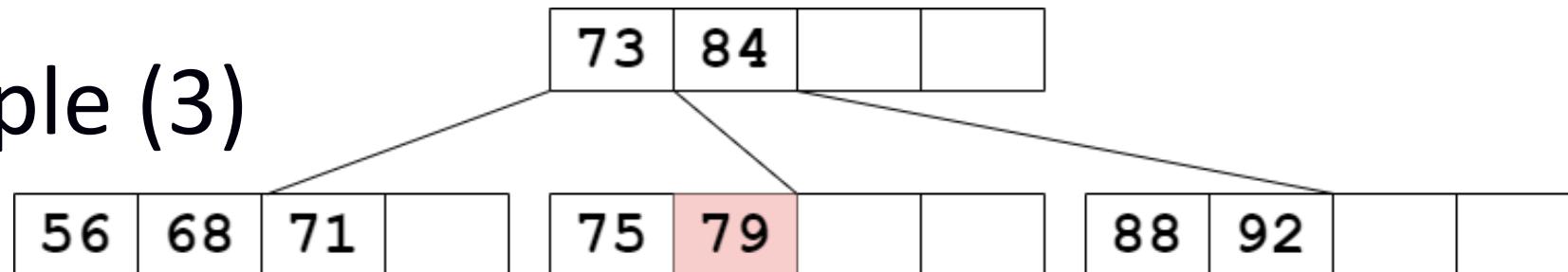
- Delete 79
- Leaf has N keys <  $\lceil m/2 \rceil - 1$
- SbL obs. case



# B-tree

## – deletion example (3)

- Delete 79
- Leaf has  $N$  keys  $< \lceil m/2 \rceil - 1$
- SbR obs. case
  - Not enough keys
- Solve root?



# B+-tree?

- All keys are in the leaf nodes
  - They contain pointers to the data
- Parent keys are found in both internal and leaf nodes
- Leaf nodes have pointers to the adjacent leaf node
  - Sequential reading

# 03 – String Based Algorithms

*Advanced Algorithms and Data Structures*



UNIVERSITY OF ZAGREB  
Faculty of Electrical  
Engineering and  
Computing

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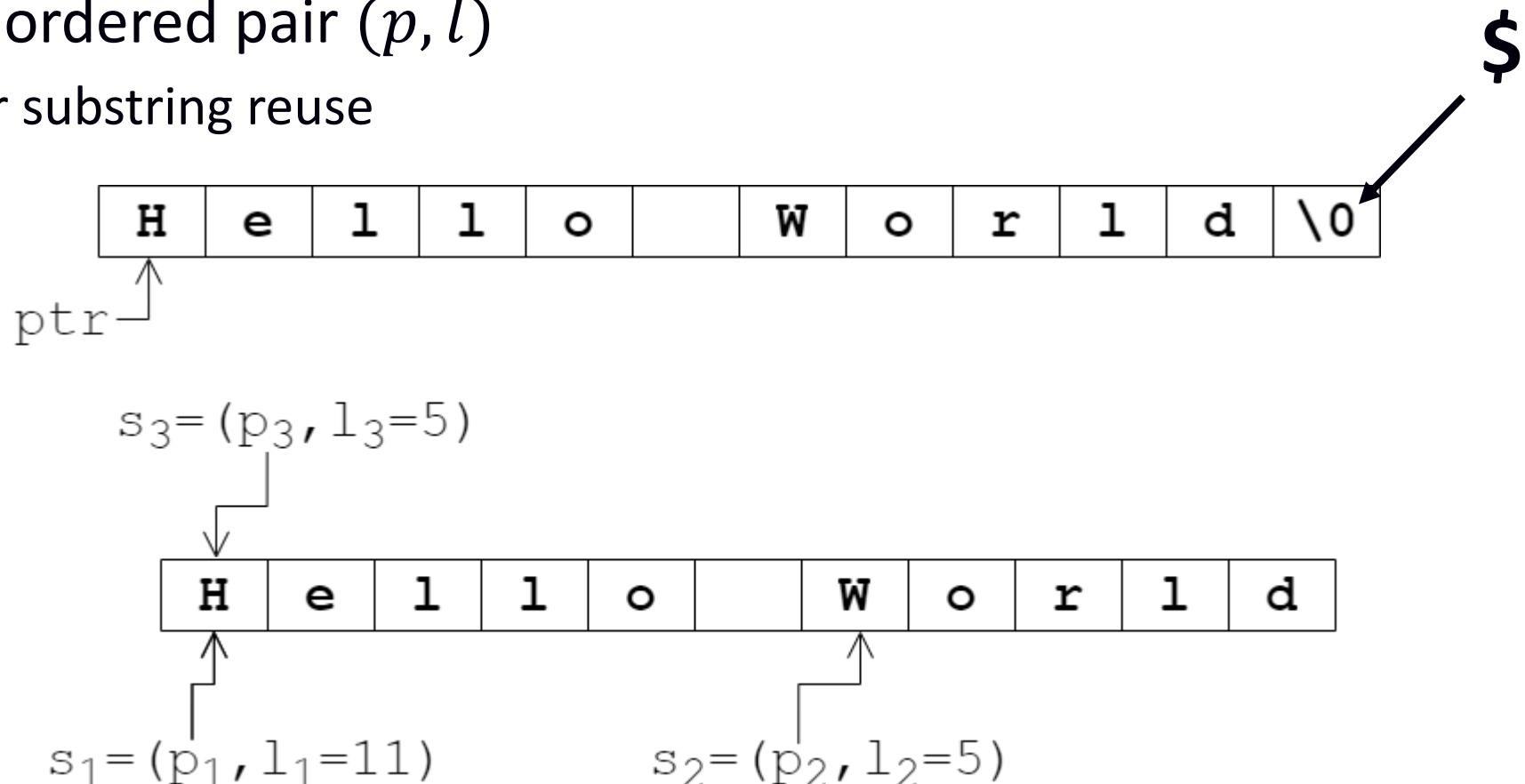


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# Representation of a string

- With a pointer and a NULL terminator
- With ordered pair  $(p, l)$ 
  - for substring reuse



# Searching a text

- Search for the existence of strings in a text corpus
  - Words in a text

$$S = \{s_i : 0 < i \leq N\}$$

- Naively – search for  $q$  in  $S$

$$O(\sum_{s_i \in S} |s_i|)$$

- Each symbol in each string

# Prefix tree – Trie (1)

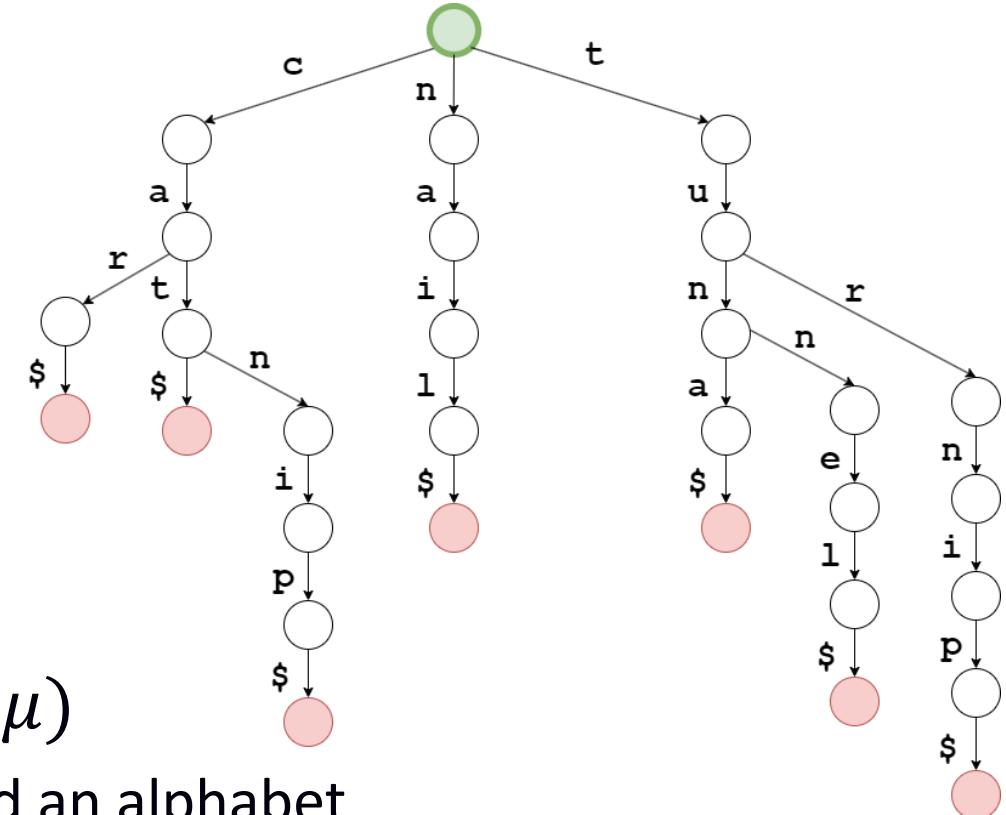
- Reads as *try*
- M-tree
- Ordered triple

$$T = (N, E, \mu)$$

With a mapping function between edges and an alphabet

$$\mu: E \rightarrow \Sigma$$

- Each node can have at most  $|\Sigma|$  Ch

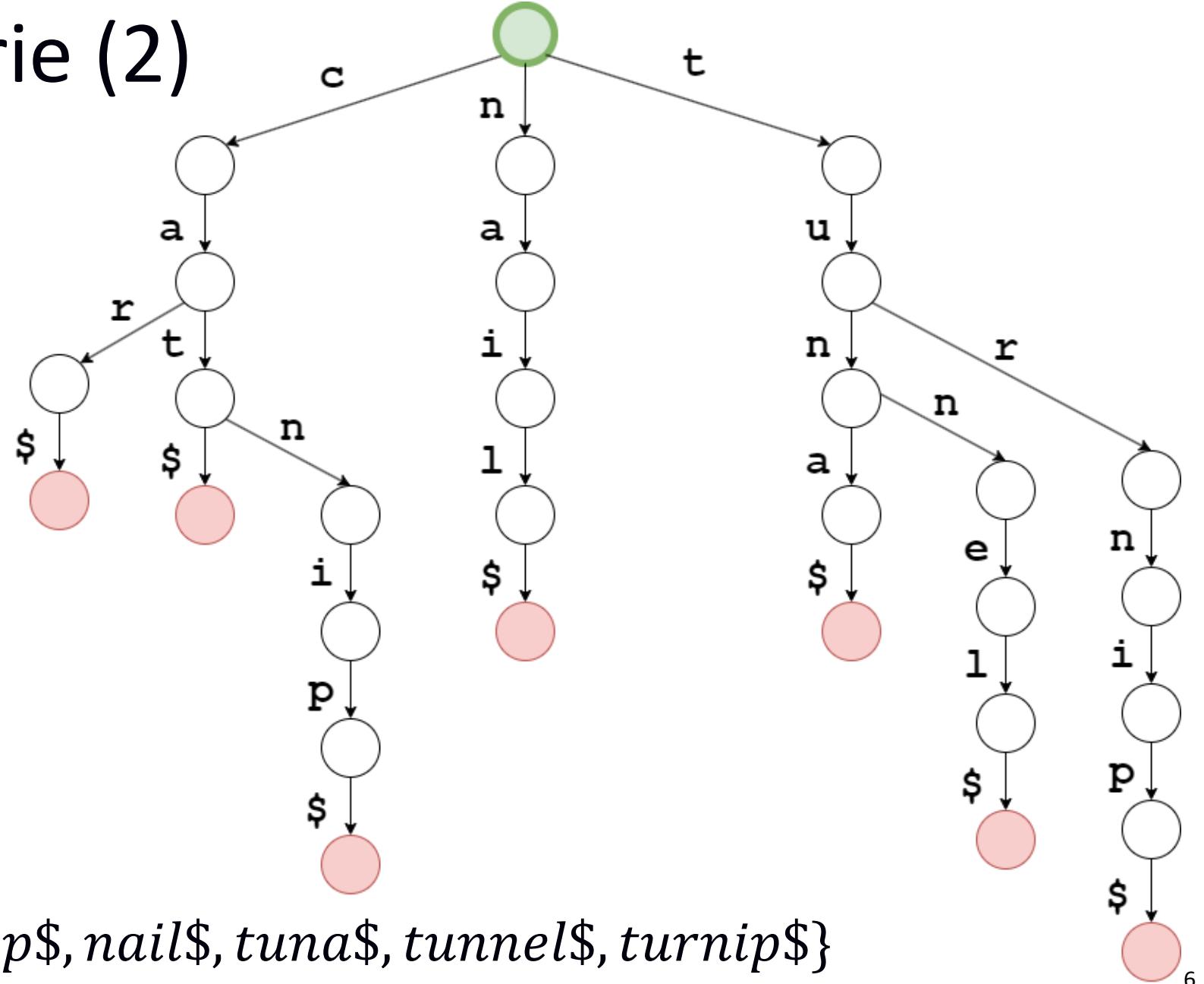


Note: Deterministic finite automata from Introduction to Theoretical Computer Science

\*with a few caveats

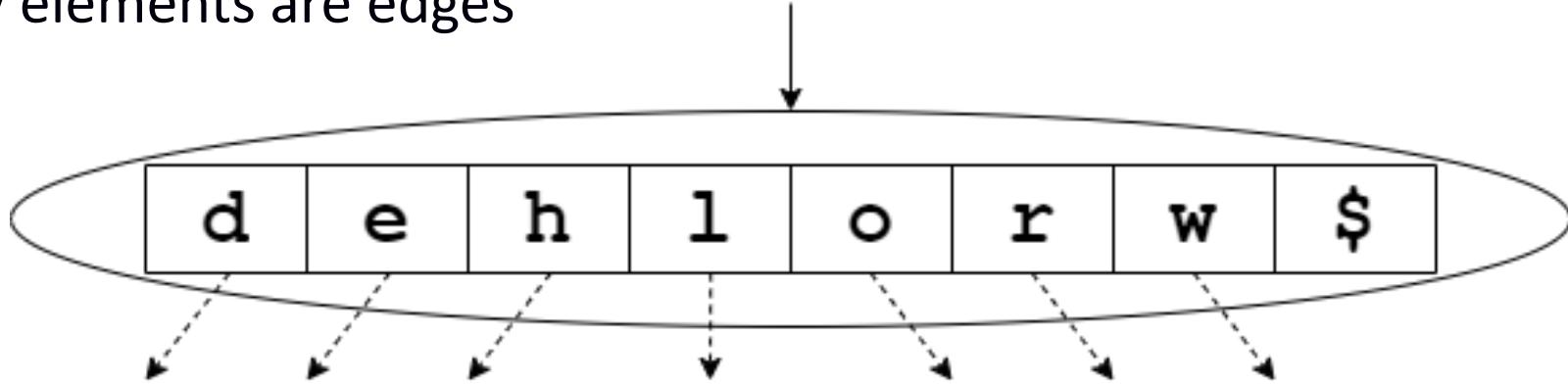
# Prefix tree – Trie (2)

- Snail?
- Reuse -nip?



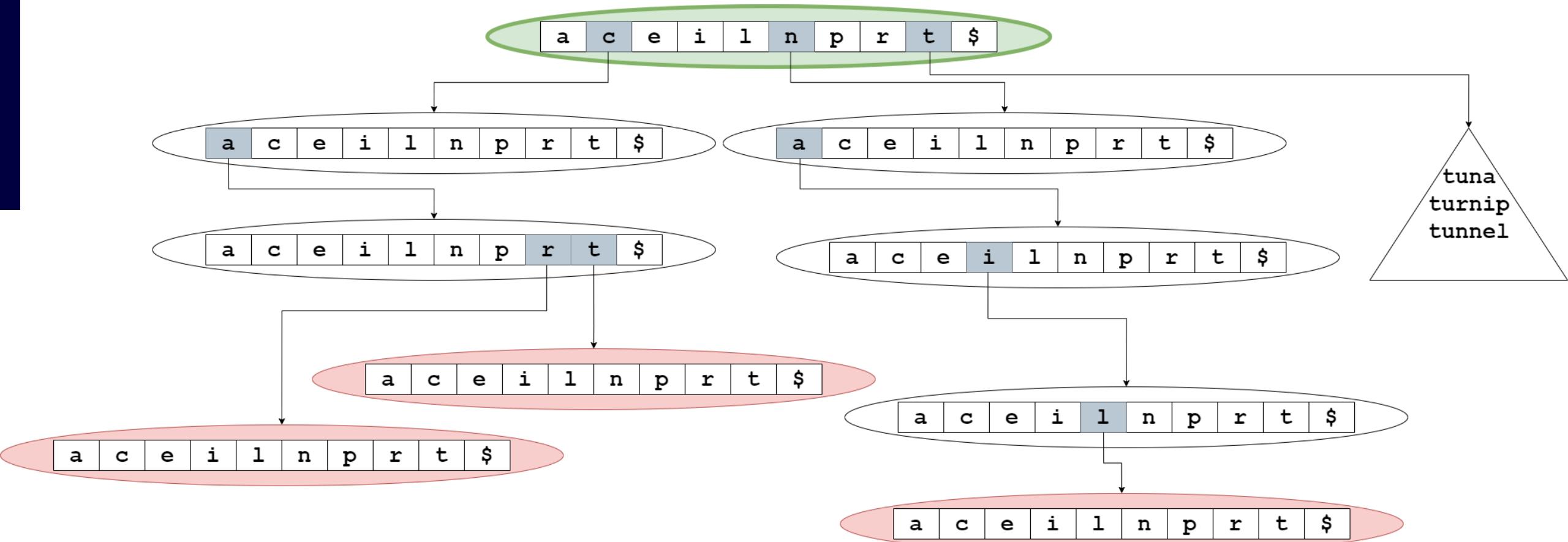
# Trie – node structure

- Array of sorted alphabet chars and a pointer for each
- Pointers lead to next char
  - Array elements are edges



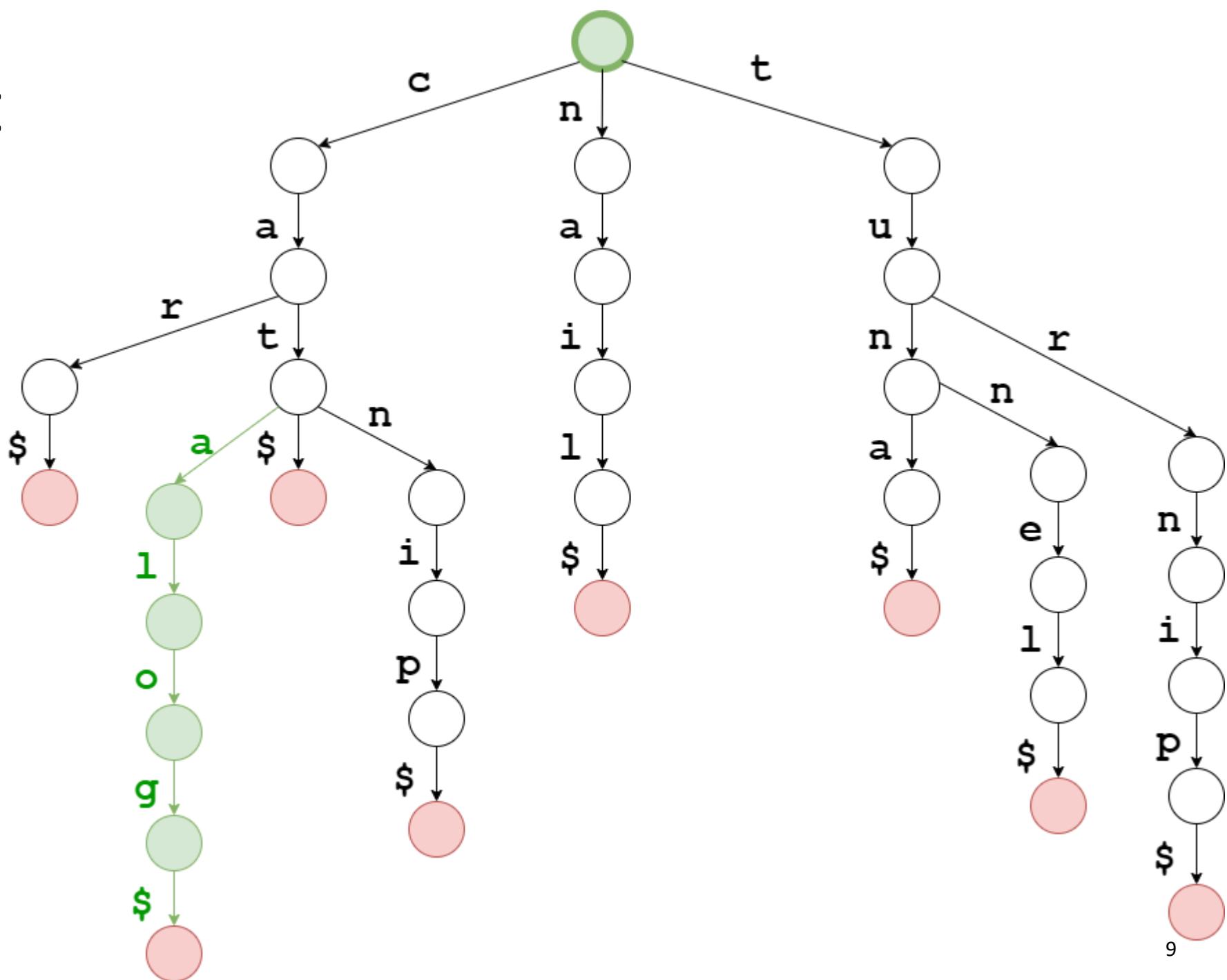
e.g. Hello World

# Trie – structured example



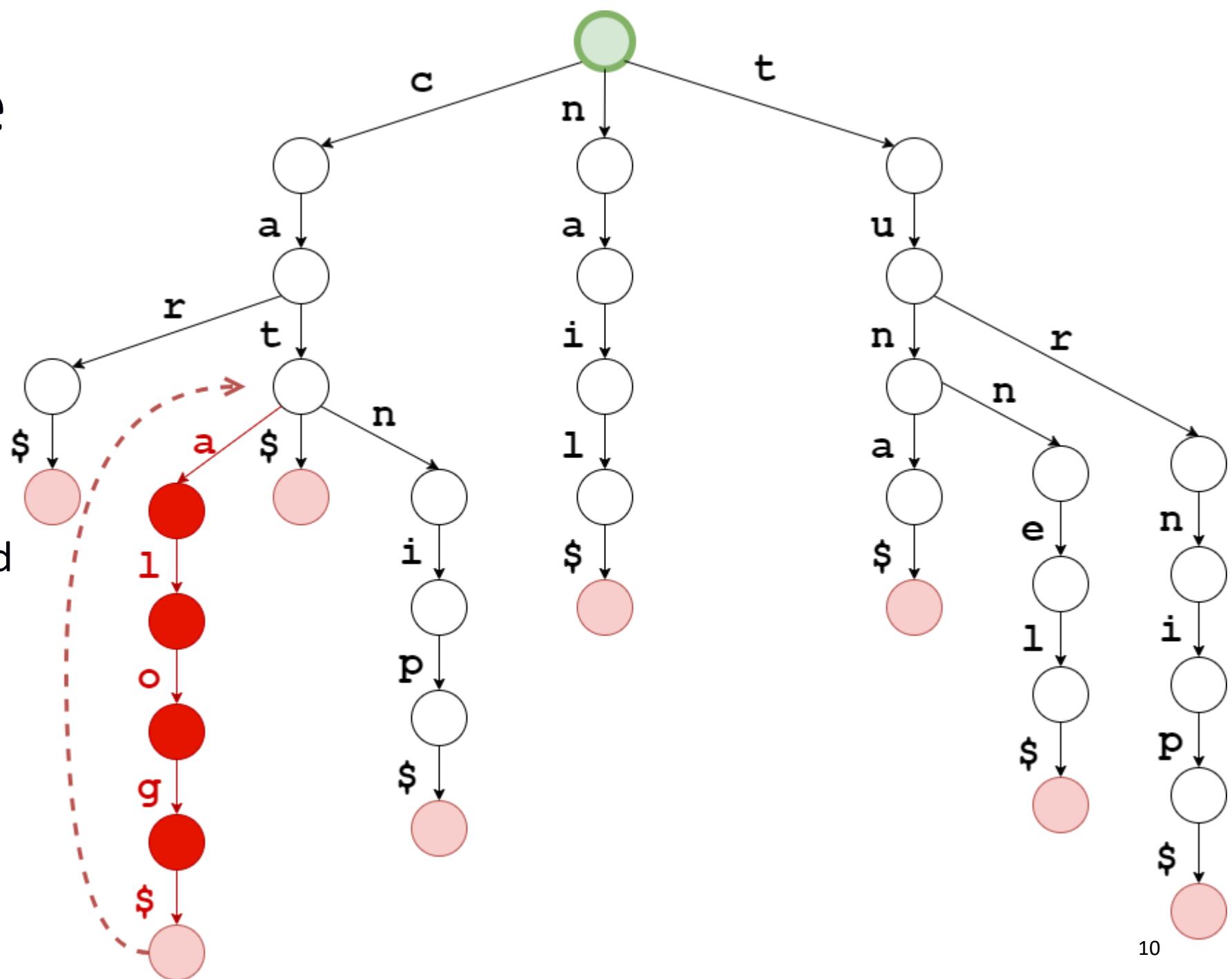
# Trie – insert

- Update alphabet
- Search until no transition
- Add children for suffix
- e.g. catalog



# Trie – delete

- Search until no transition
- Remove backwards
  - While single child
- e.g. catalog

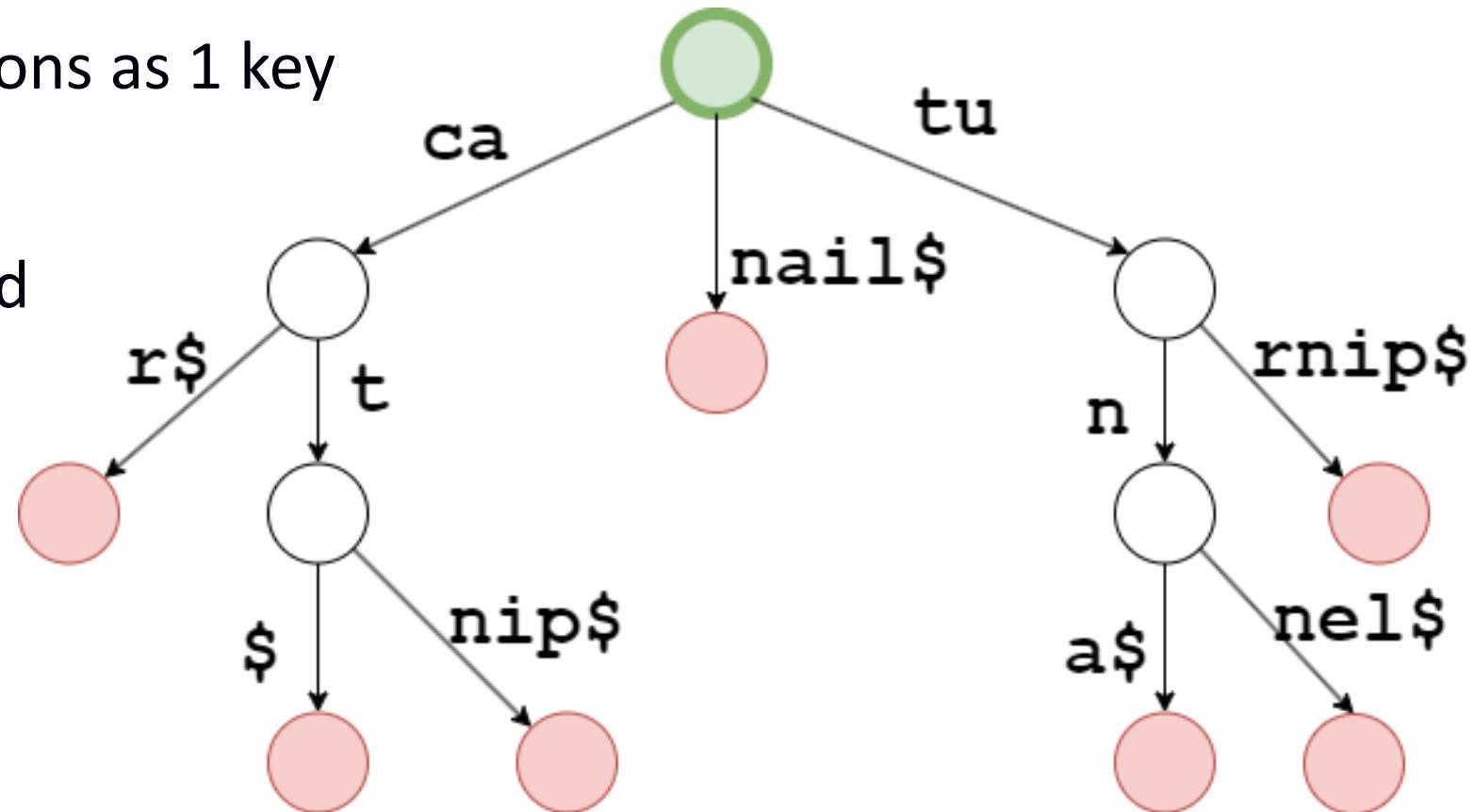


# Trie – complexity

- Space:  $O(\sum_{s_i \in S} |s_i| * |\Sigma|)$ 
  - All letters for each word
- Search time:  $O(|q|)$ 
  - Length of the query string
- A tradeoff for including the array
- Insert:  $O(|q|)$ 
  - Find prefix and complete for suffix
- Delete:  $O(|q|)$ 
  - Find terminal for the word and rollup the unique branch
    - Only child or one of...

# Patricia tree (trie?)

- A compressed Trie
- A sequence of transitions as 1 key
- Less memory is utilized

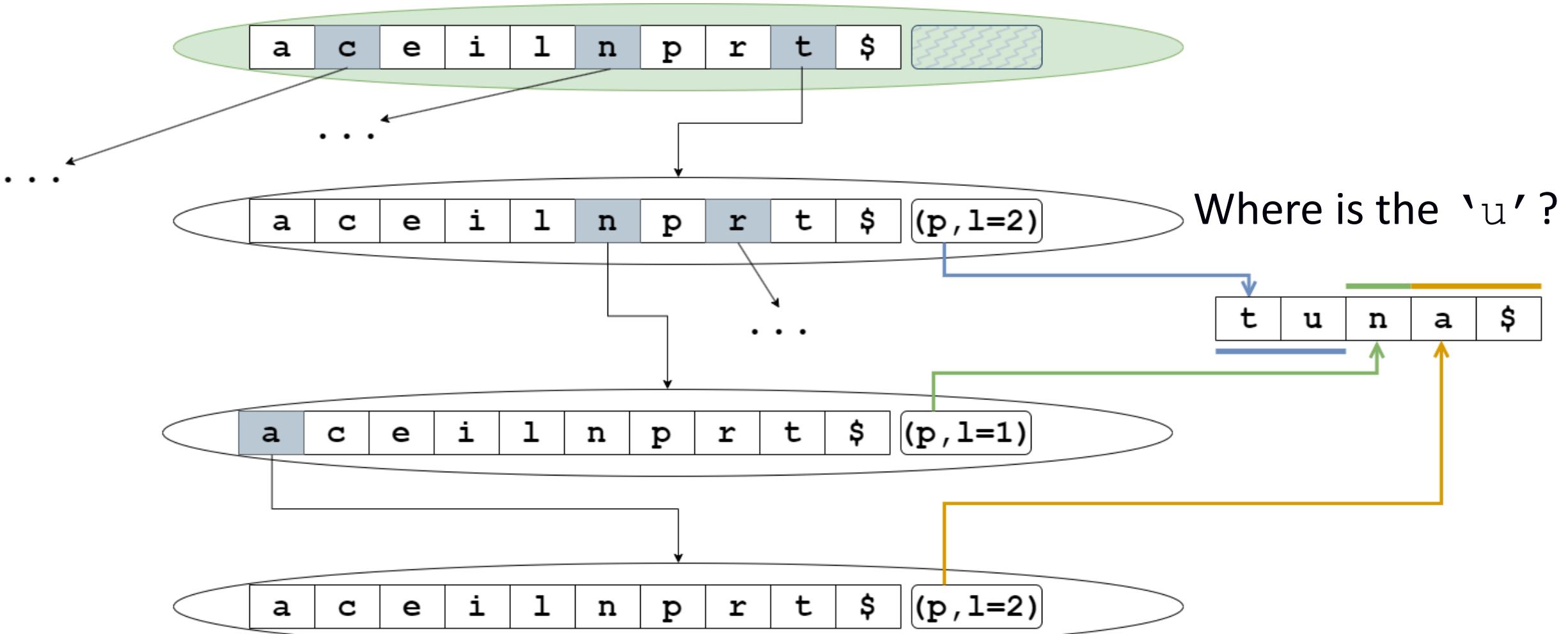


# Patricia tree – node structure (1)

- The nodes are references that contain:
  - $(p, l)$  to an instance of a string
    - First created
    - Pointers to the Ch substrings
  - Internal nodes have at least 2 Ch

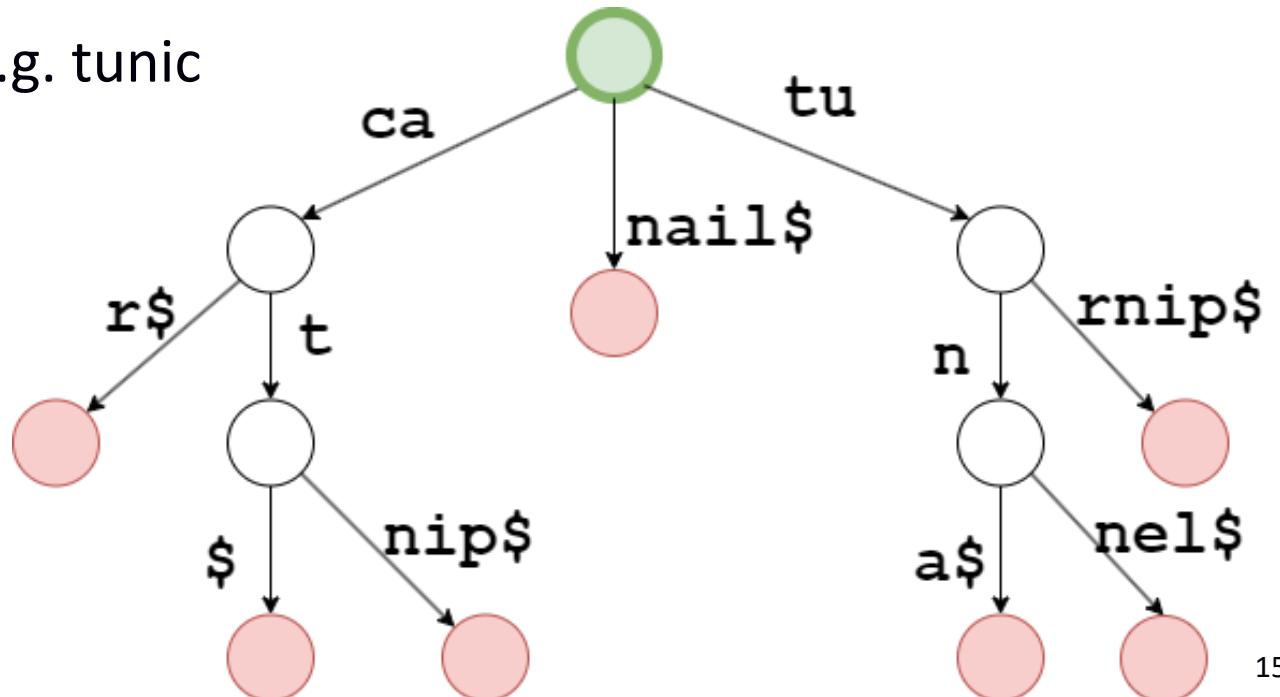
Note: „references” as in index nodes of a database

# Patricia tree – node structure (2)



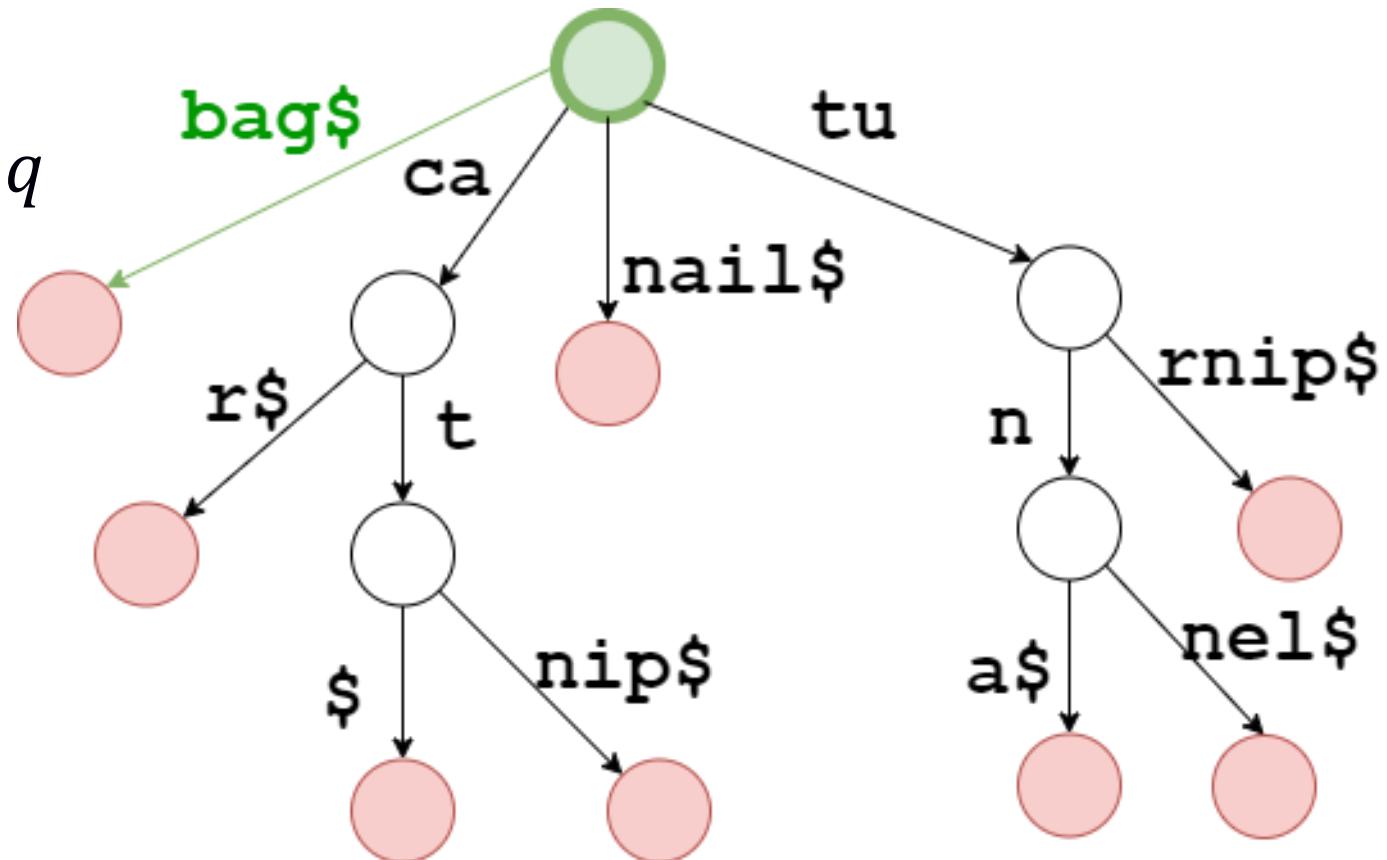
# Patricia tree – insert (1)

- Starts with search
- 3 scenarios:
  1. No transition found – e.g. bag
  2. Partial transition found – e.g. name
  3. Complete transition found – e.g. tunic
- Adds string to memory
- Complexity:  $O(|s| + |\Sigma|)$ 
  - New word after entire alphabet is matched



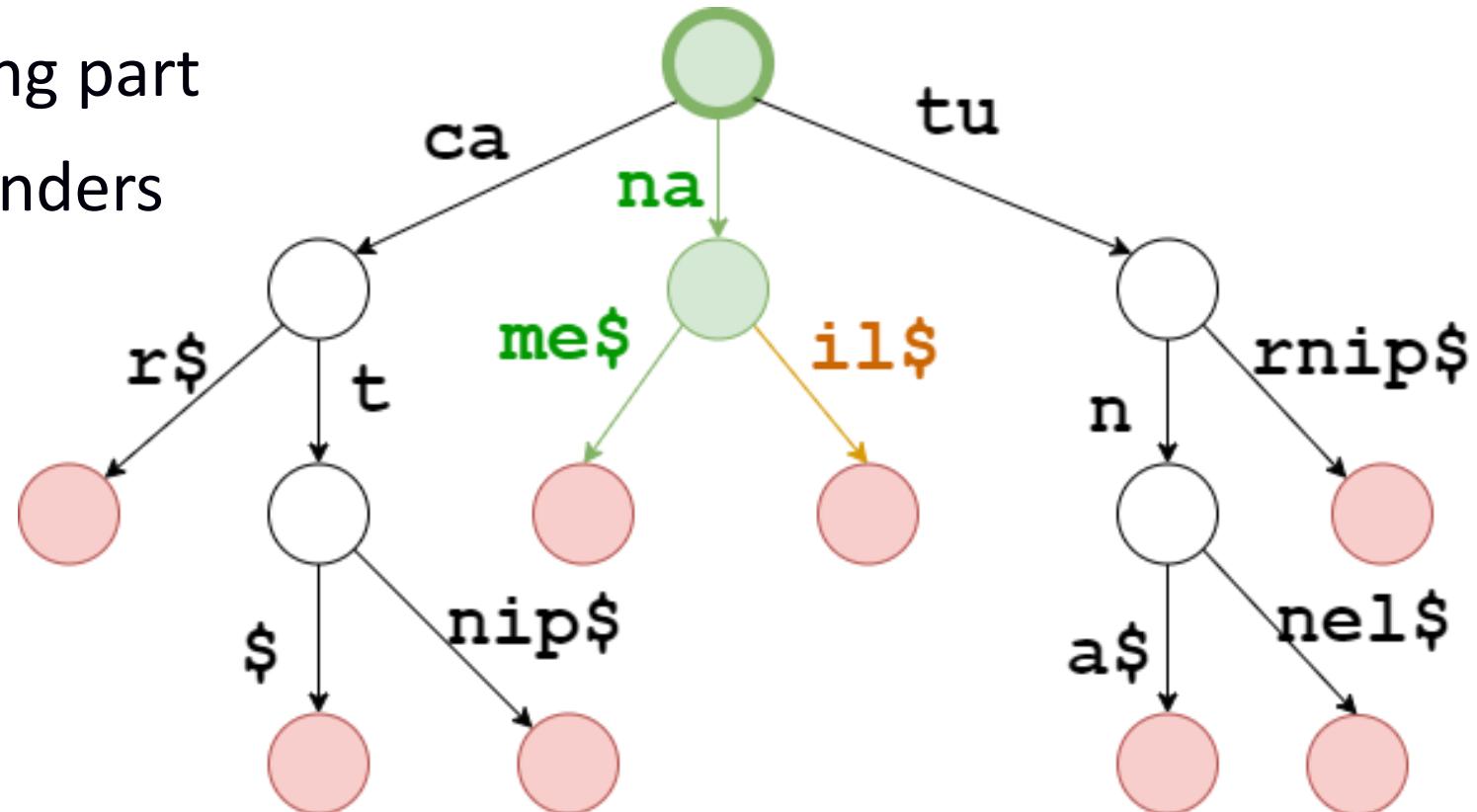
# Patricia tree – insert (2)

- No transition
- Create new node with entire  $q$
- “bag”



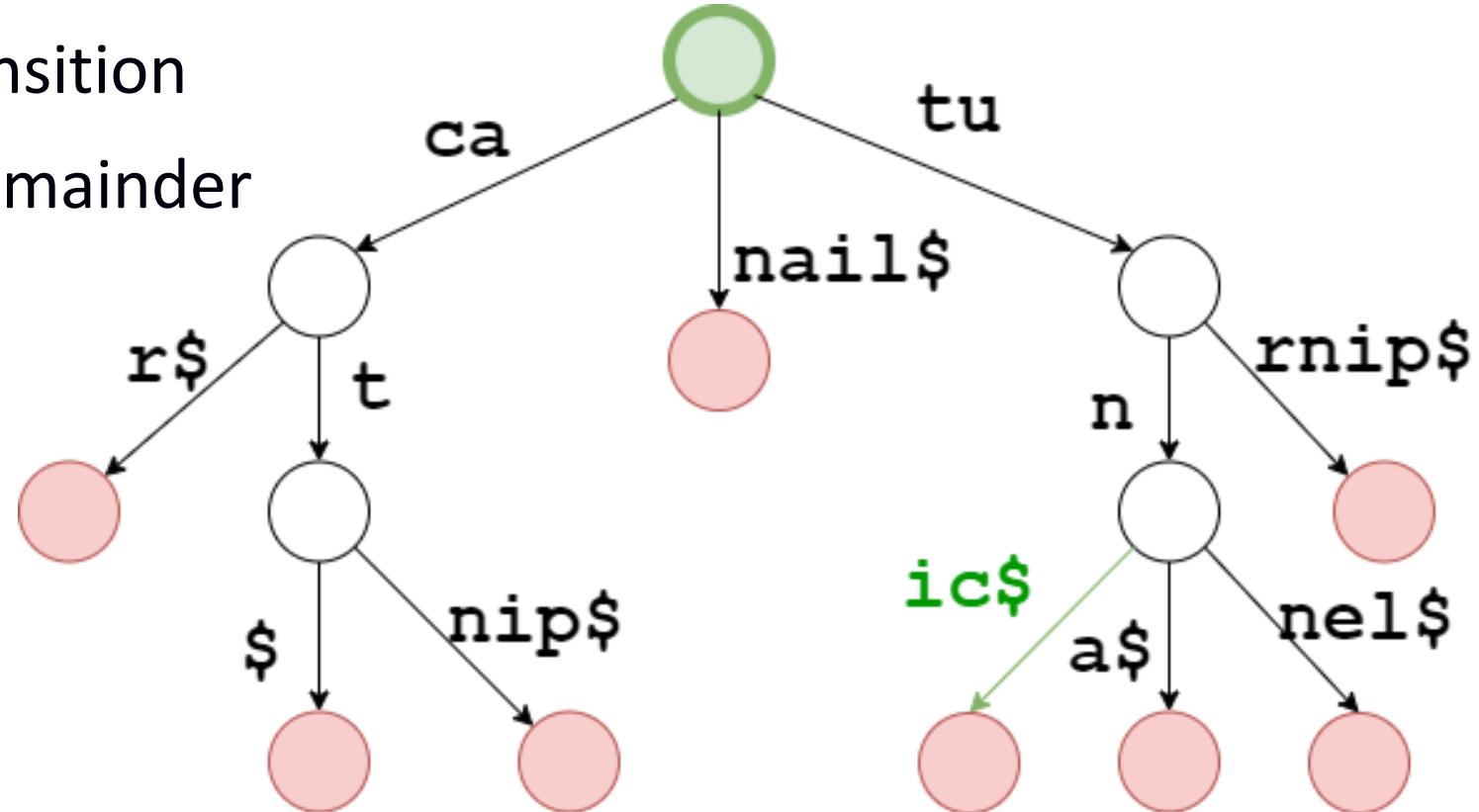
# Patricia tree – insert (3)

- Partial transition
- Create Pr with the matching part
- Create Chs with the remainders
- “name”



# Patricia tree – insert (4)

- Full transition
- Find the last matching transition
- Create new Ch with the remainder
- “tunic”

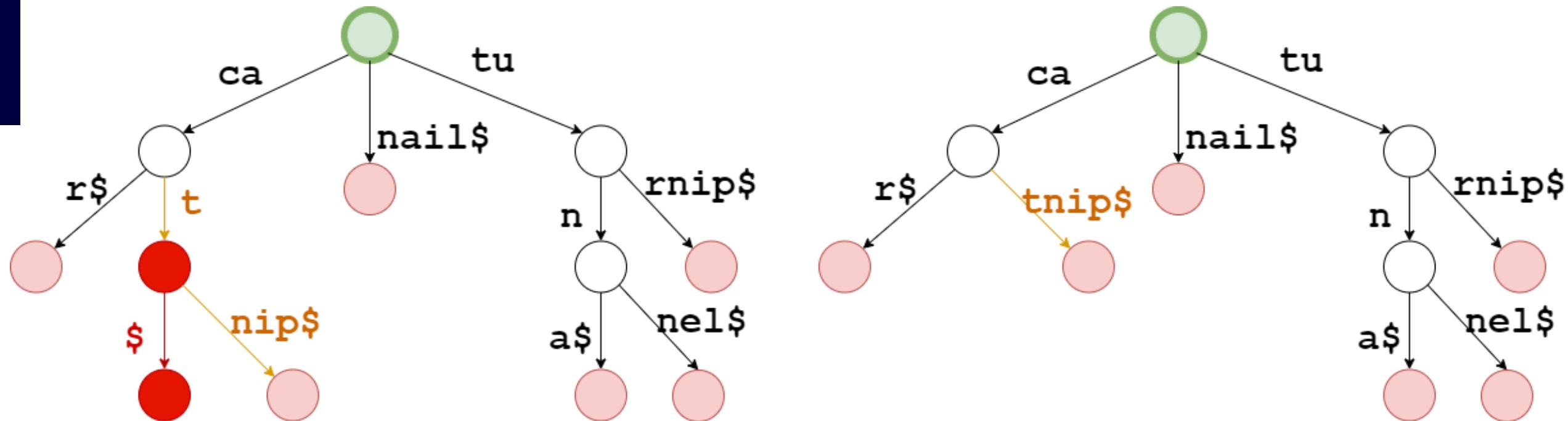


# Patricia tree – delete (1)

- Search
- Remove leaf nodes that match
- Compress internal nodes that have one child after deletion
- Remove string from memory

# Patricia tree – delete (2)

- Delete “cat”



# Prefix trees in general

- Use in:
  - Associative arrays – e.g. dictionaries
    - Haskell
  - Sorting
  - IP routing
  - Inverted indexes
  - Proofs about inductive data structures or defining recursive functions
    - Coq
  - Symbol tables
    - Lexing

# Suffix arrays

# Suffix trees

# Array suffix

- All suffixes of an array
- e.g. program\$:
  - program\$
  - rogram\$
  - ogram\$
  - gram\$
  - ram\$
  - am\$
  - m\$
  - \$
- Best way to store it?
  - Naively, a lot of copies
  - $(p, l) \rightarrow |s|$  ordered pairs

# Suffix array (1)

- A sorted array of suffixes

$i$	$A_i$	$t_{A_i}$						
1	8	\$						
2	6	a	m	\$				
3	4	g	r	a	m	\$		
4	7	m	\$					
5	3	o	g	r	a	m	\$	
6	1	p	r	o	g	r	a	m
7	5	r	a	m	\$			
8	2	r	o	g	r	a	m	\$

- Creation is costly:  $O(n^2 \log n)$ 
  - Sort  $O(n \log n)$  and char comparison  $O(n)$  in sorting

# Suffix array (2)

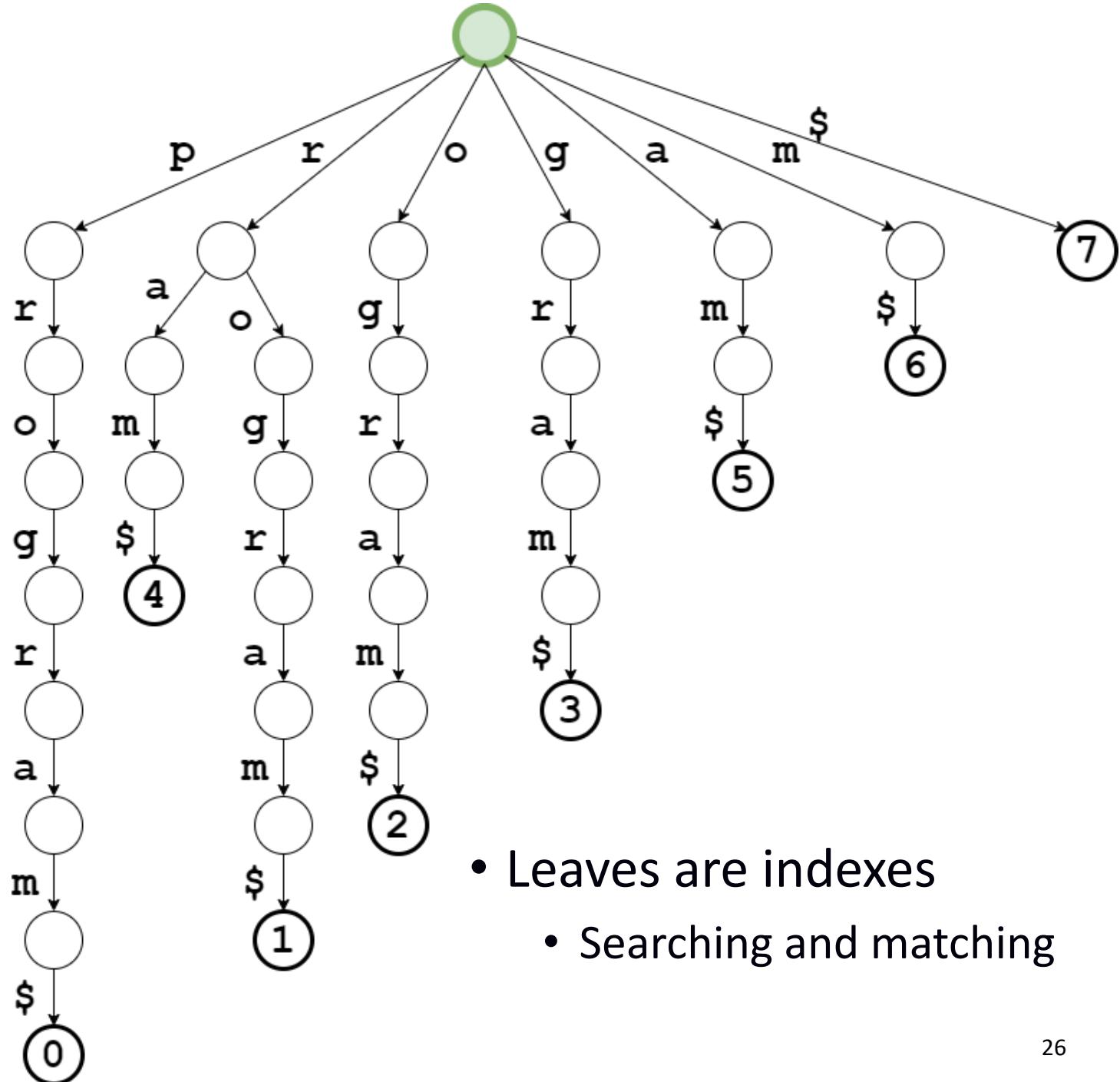
- Goal: Check for the existence of a  $q$  in  $a$  in a string  $t$
- Search naively  $O(|q| * n)$
- Search with binary alg.  $O(|q| \log n)$

\*Enhancements – AADS script

# Suffix trie (1)

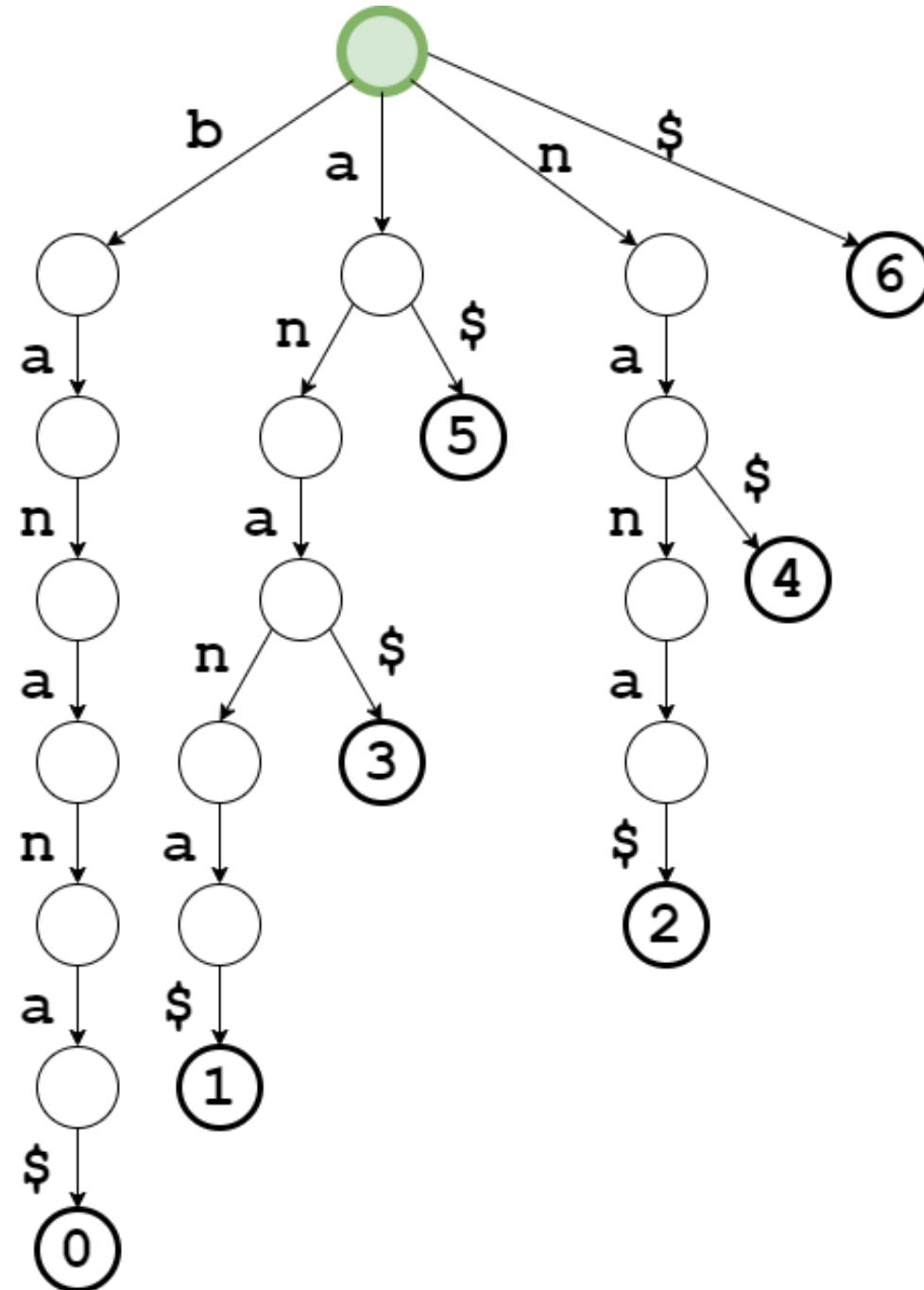
- What if we use the Trie to index the suffix array?

p	r	o	g	r	a	m	\$
r	o	g	r	a	m	\$	
o	g	r	a	m	\$		
g	r	a	m	\$			
r	a	m	\$				
a	m	\$					
m	\$						
\$							



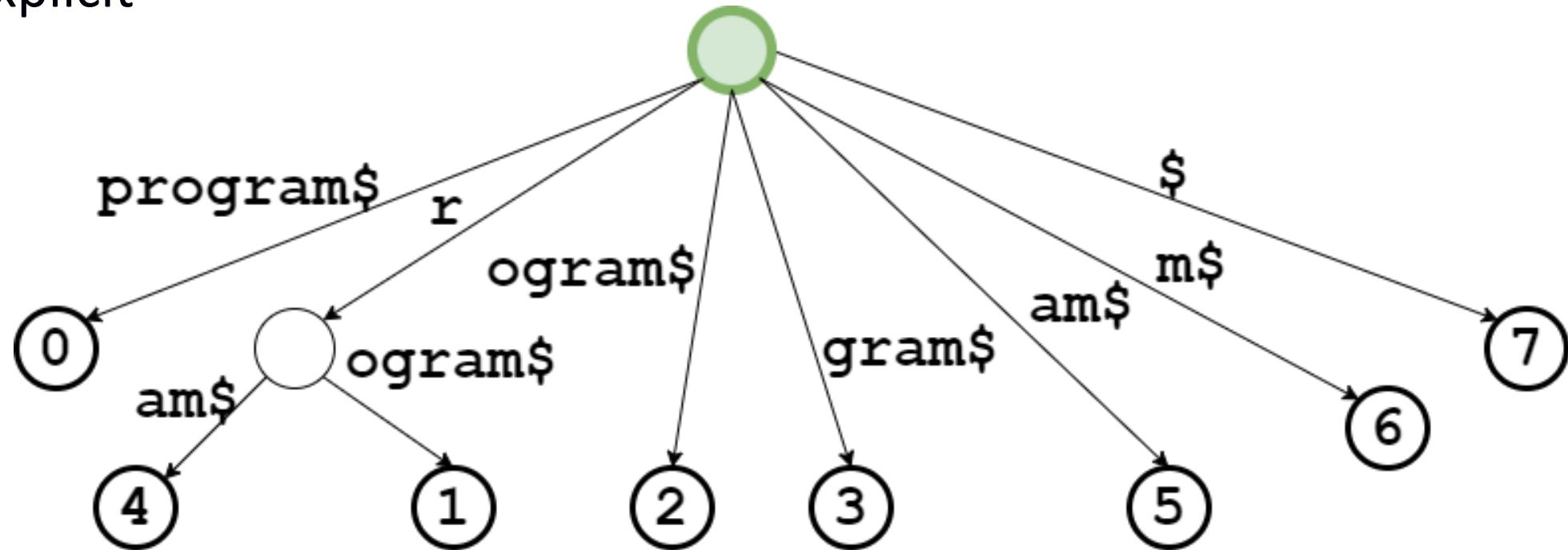
# Suffix trie (2)

- A more repetitive example



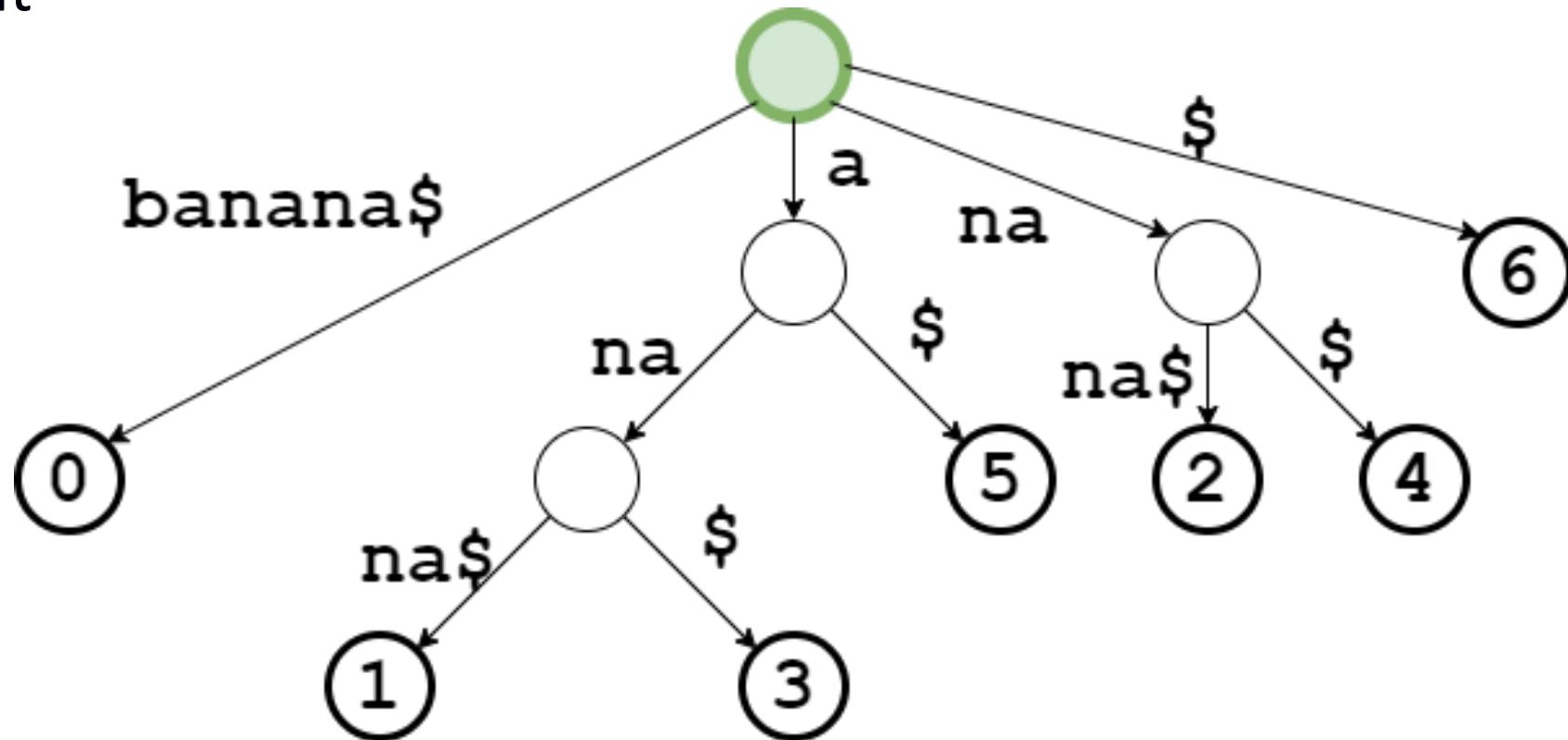
# Compressed suffix trie – Patricia tree (1)

- Explicit



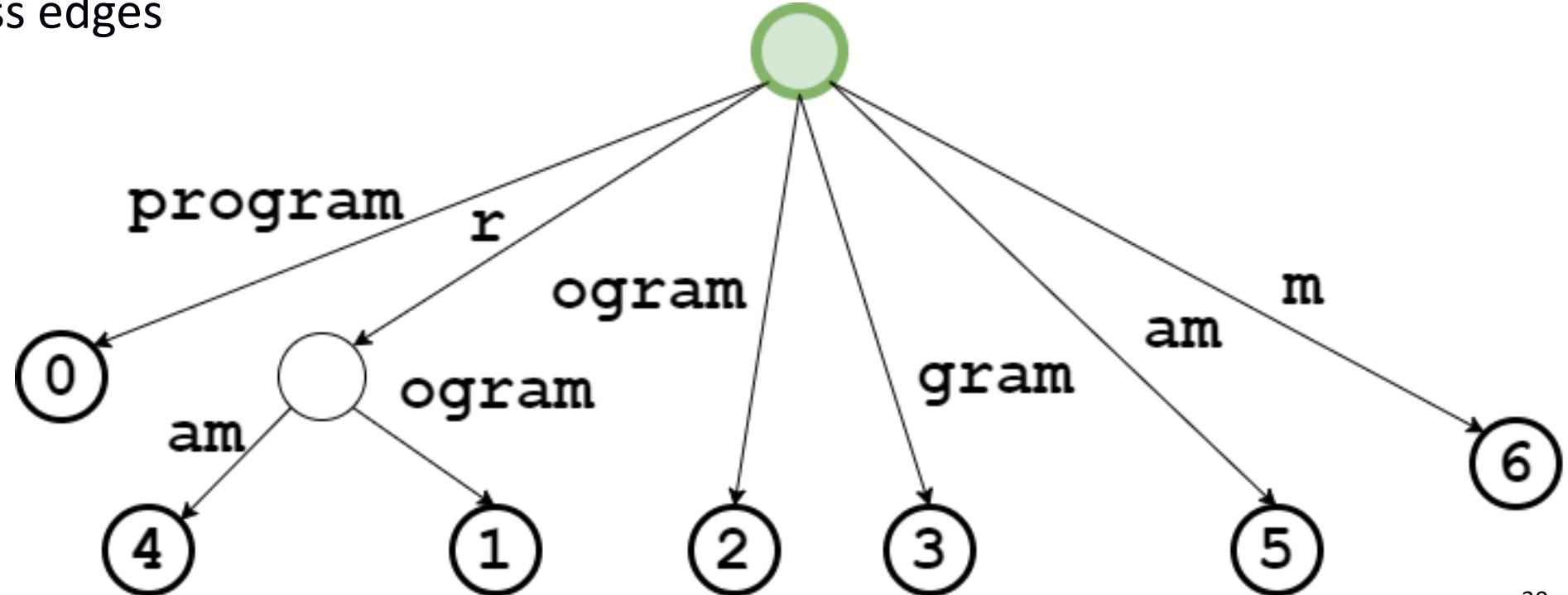
# Compressed suffix trie – Patricia tree (2)

- **Explicit**



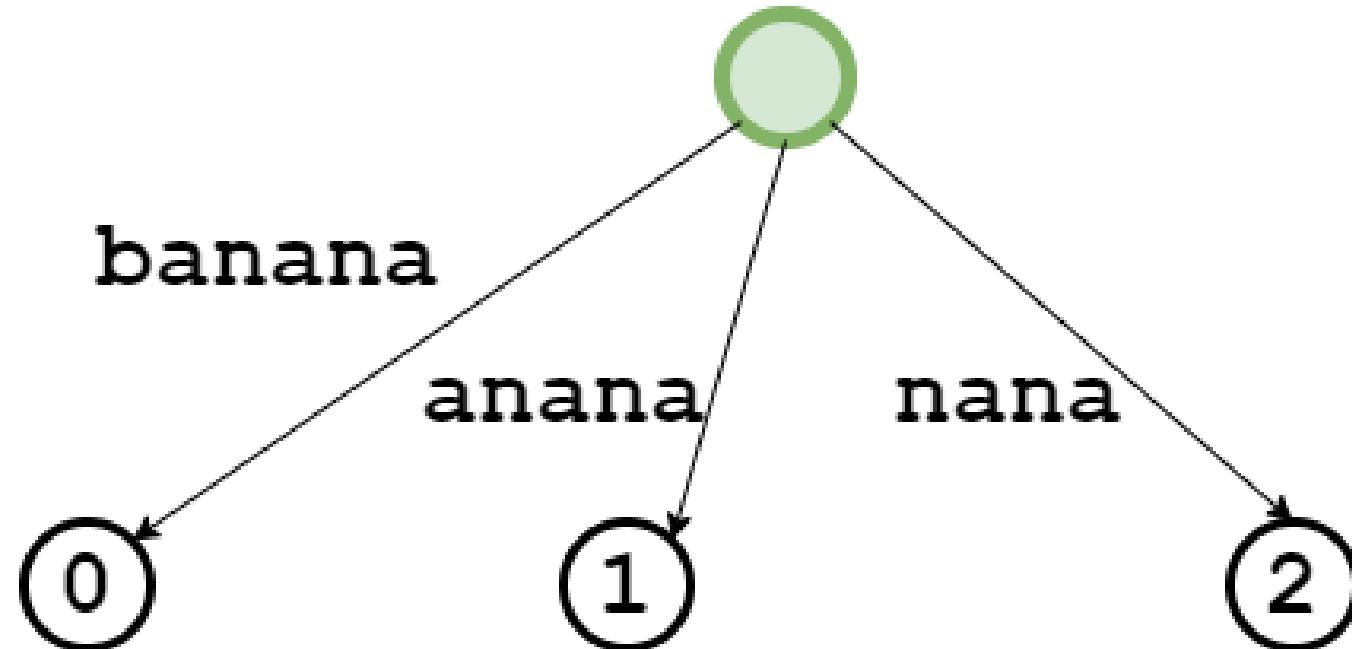
# Compressed suffix trie – Patricia tree (3)

- Implicit suffix tree:
  1. Remove all terminators from edges
  2. Remove all empty edges; including subtrees
  3. Compress edges



# Compressed suffix trie – Patricia tree (4)

- Implicit suffix tree:
  1. Remove all terminators from edges
  2. Remove all empty edges; including subtrees
  3. Compress edges



# Ukkonen's algorithm (1)

- Online creation of implicit compressed suffix tries
- Creation:  $O(|S|)$
- Contains steps and substeps
  - Steps analyze each prefix ( $i \in [0, |S| - 1]$ )
  - Substeps analyze each suffix of a prefix ( $j \in [0, i + 1]$ )
- Each step we add  $S[i + 1]$ 
  - There is a step  $i = -1$  where we just add  $S[0]$  to the tree
- Each substep we use  $\beta = S[j..i]$  to traverse the existing tree

# Ukkonen's algorithm (2)

- 3 cases when adding:

**C1:**  $\beta$  traverses to a leaf

-> Add  $S[i+1]$  to the leaf edge

**C2:**  $\beta$  stopped mid-traversal AND the next char is not  $S[i+1]$

-> Split at stop (create splitting node)

-> Add leaf edge with  $S[i+1]$

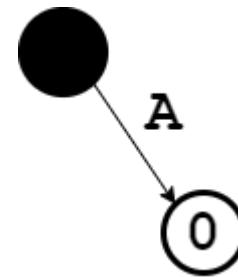
-> Add current  $j$  as leaf

**C3:**  $\beta$  stopped mid-traversal AND the next char is  $S[i+1]$

-> Do nothing

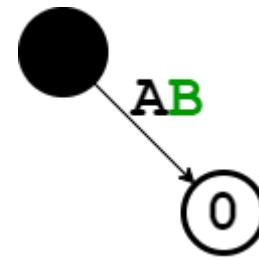
# Ukkonen's algorithm – example (1)

Step -1	S=ABABABC\$	



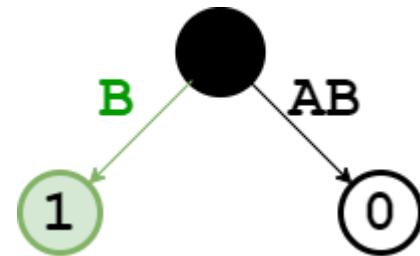
# Ukkonen's algorithm – example (2)

Step i=0	S=ABABABC\$	S[i+1]=B
j=0	$\beta=S[j..i]=S[0..0]=A$	Case: C1



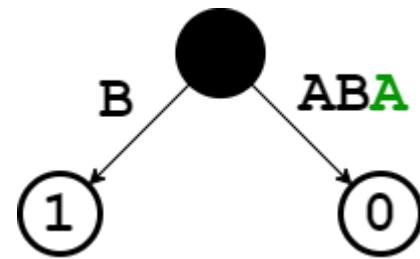
# Ukkonen's algorithm – example (3)

Step i=0	S=ABABABC\$	S[i+1]=B
j=1	$\beta=S[j..i]=S[1..0]=\epsilon$	Case: C2



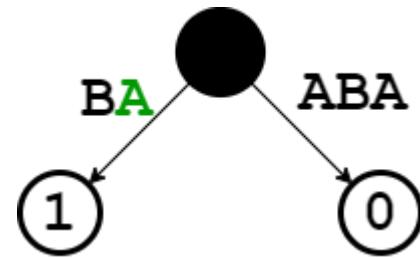
# Ukkonen's algorithm – example (4)

Step i=1	S=ABABABC\$	S[i+1]=A
j=0	$\beta=S[j..i]=S[0..1]=AB$	Case: C1



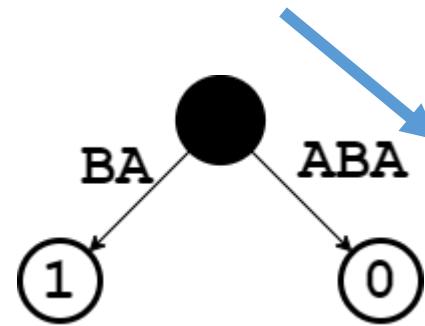
# Ukkonen's algorithm – example (5)

Step i=1	S=ABABABC\$	S[i+1]=A
j=1	$\beta=S[j..i]=S[1..1]=B$	Case: C1



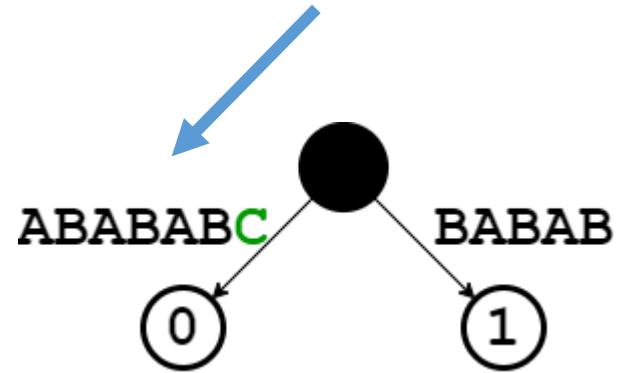
# Ukkonen's algorithm – example (6)

Step i=1	S=ABABABC\$	S[i+1]=A
j=2	$\beta=S[j..i]=S[2..1]=\epsilon$	Case: C3



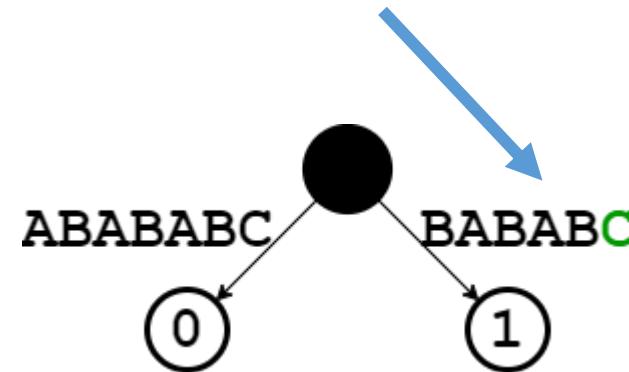
# Ukkonen's algorithm – example (7)

Step i=5	S=ABABABC\$	S[i+1]=C
j=0	$\beta=S[j..i]=S[0..5]=ABABAB$	Case: C1



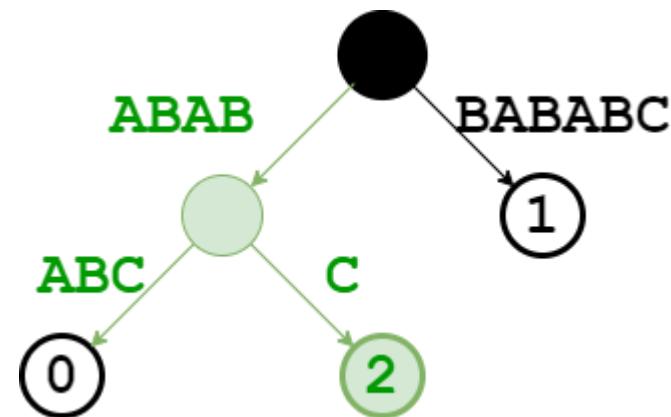
# Ukkonen's algorithm – example (8)

Step i=5	S=ABABABC\$	S[i+1]=C
j=1	$\beta=S[j..i]=S[1..5]=BABAB$	Case: C1



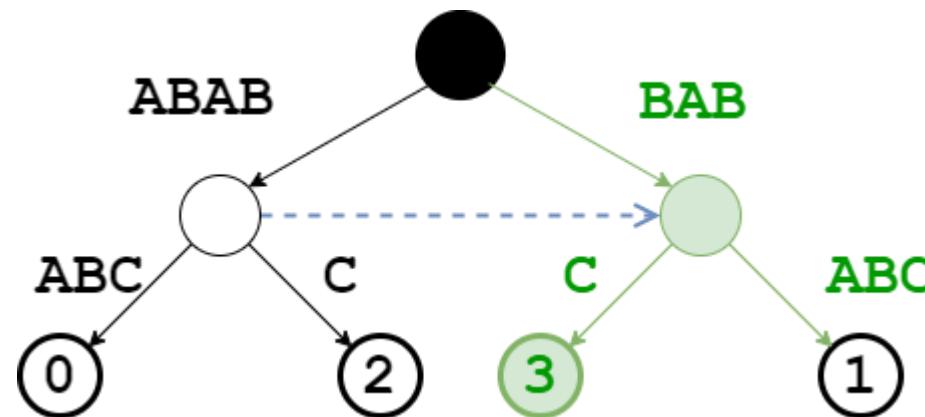
# Ukkonen's algorithm – example (9)

Step i=5	S=ABABABC\$	S[i+1]=C
j=2	$\beta=S[j..i]=S[2..5]=ABAB$	Case: C2



# Ukkonen's algorithm – example (10)

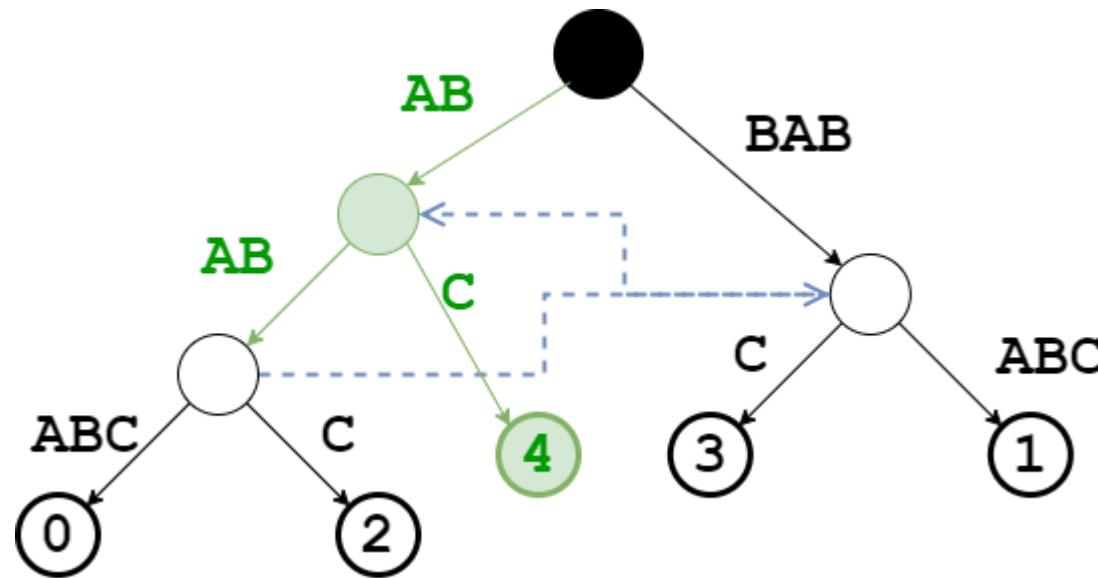
Step i=5	S=ABABABC\$	S[i+1]=C
j=3	$\beta=S[j..i]=S[3..5]=BAB$	Case: C2



**Suffix links** are added only on C2 when an internal node is added  
-> Link toward and from the created node with  $X\alpha \rightarrow \alpha$

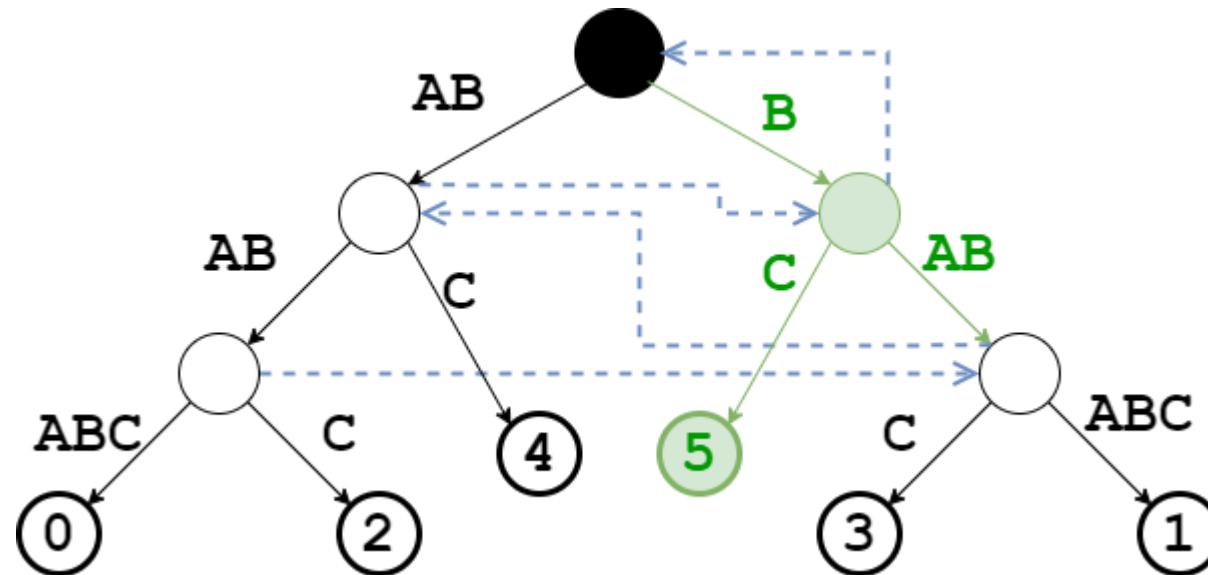
# Ukkonen's algorithm – example (11)

Step i=5	S=ABABABC\$	S[i+1]=C
j=4	$\beta=S[j..i]=S[4..5]=AB$	Case: C2



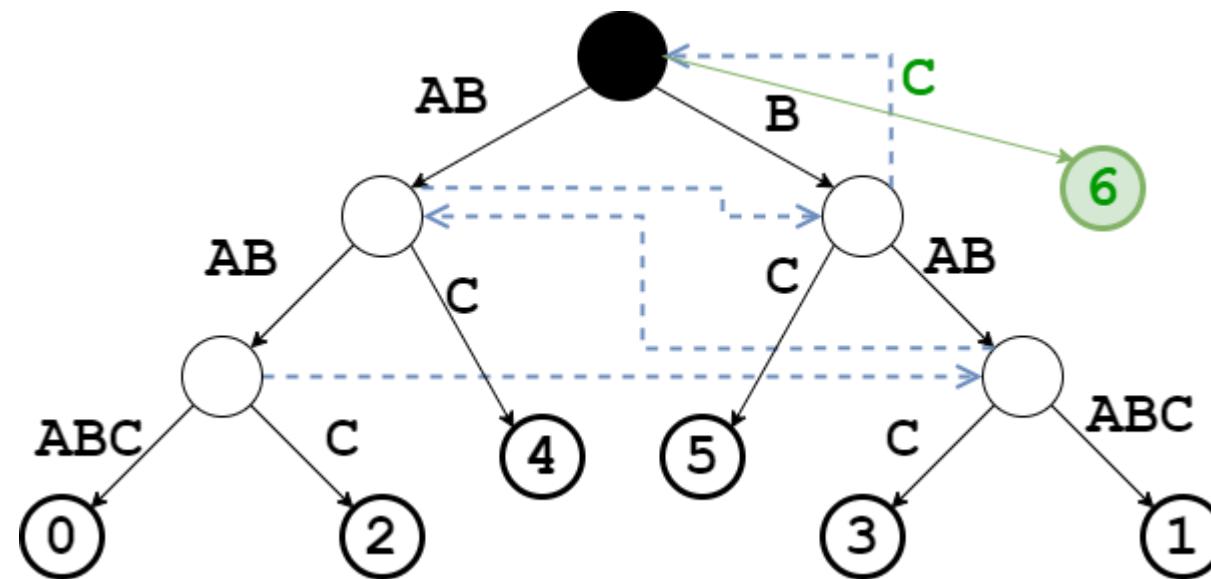
# Ukkonen's algorithm – example (12)

Step i=5	S=ABABABC\$	S[i+1]=C
j=5	$\beta=S[j..i]=S[5..5]=B$	Case: C2



# Ukkonen's algorithm – example (13)

Step i=5	S=ABABABC\$	S[i+1]=C
j=6	$\beta=S[j..i]=S[6..5]=\epsilon$	Case: C2



# Trees aren't your thing?

- <https://cp-algorithms.com/string/suffix-automaton.html>

# 04 – Geometric algorithms

*Advanced Algorithms and Data Structures*



UNIVERSITY OF ZAGREB  
Faculty of Electrical  
Engineering and  
Computing

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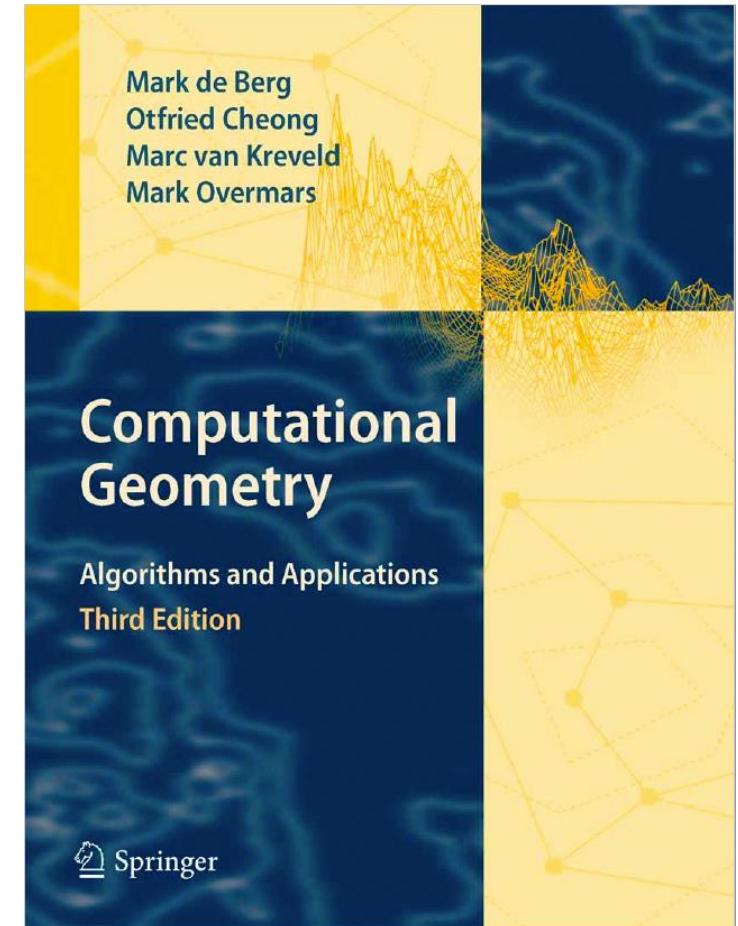


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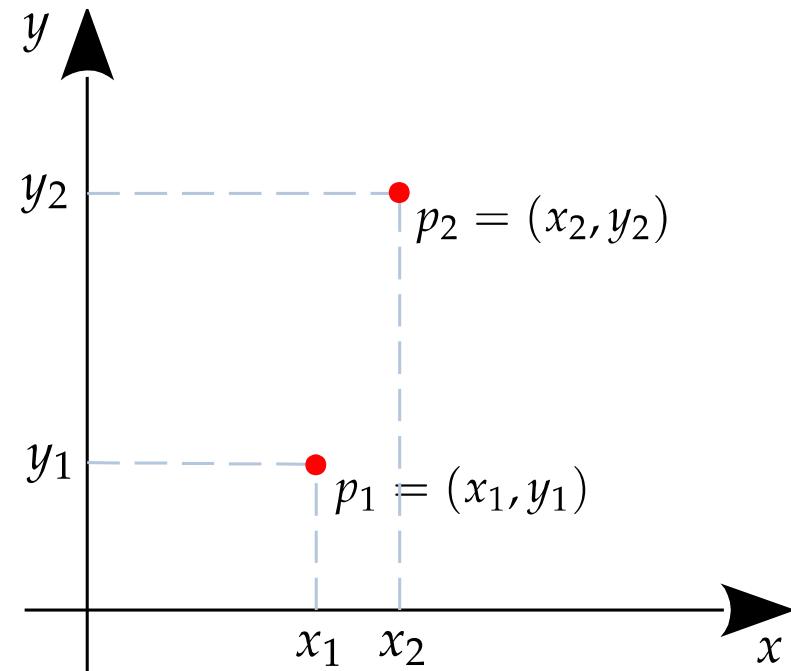
# Geometric algorithms

- Uses in:
  - Computer graphics
  - Spatial product design (CAD)
  - Robotics
  - Geographic information systems
- Not just 2D and 3D
  - Consider everything as dimensionally expandable
  - $\mathbb{R} \times \mathbb{R} \times \mathbb{R} \times \dots \times \mathbb{R}$



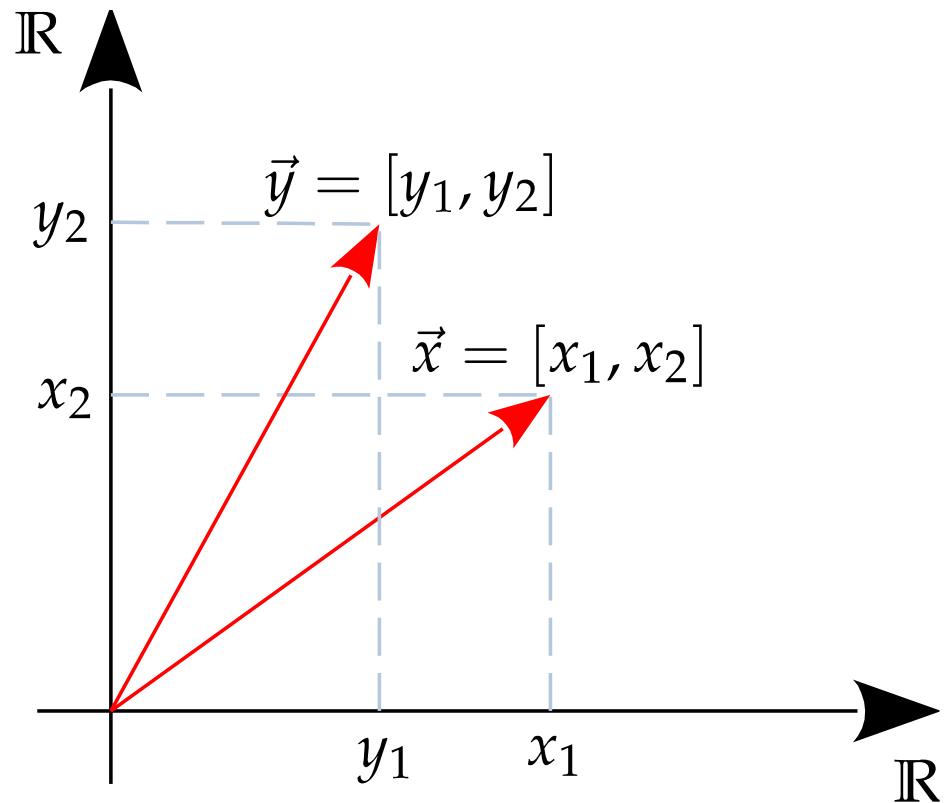
# Basic concepts – point (1)

- Point
  - Infinitesimal element in  $\mathbb{R}^n$
  - Cartesian coordinate system



# Basic concepts – point (2)

- For multiple dimensions
- As a vector:  $\vec{x} = [x_1 \ x_2 \ x_3 \dots \ x_n]$
- Dimensions as axes
  - Now numbered 1, 2



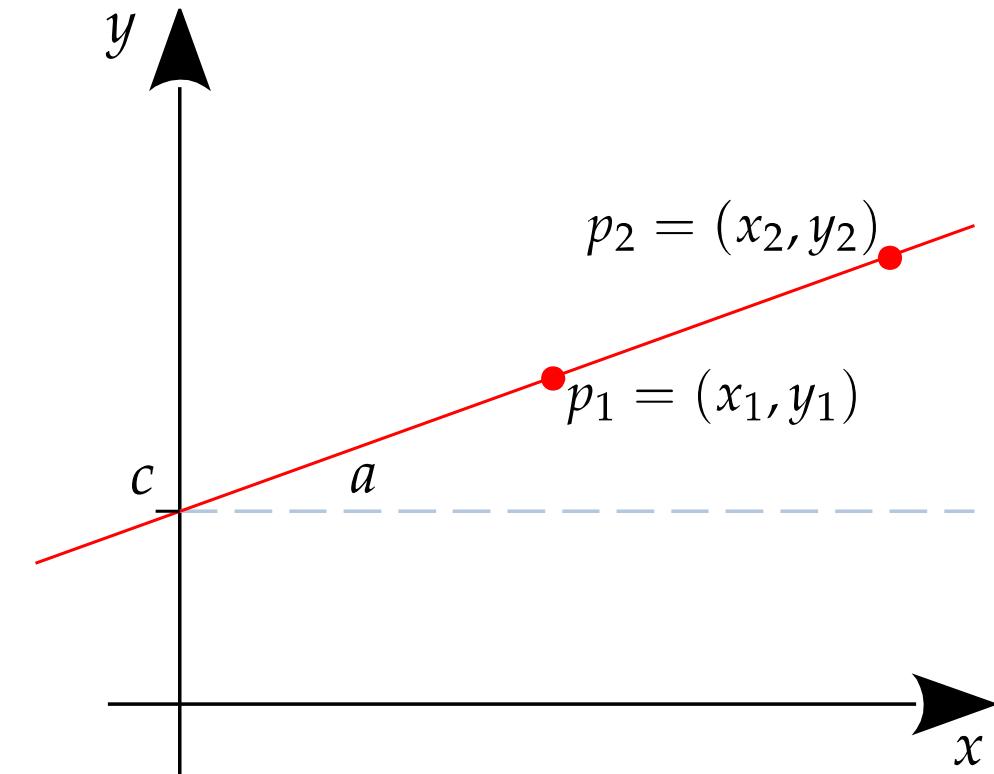
# Basic concepts - line

- Infinite amount of points:

$$L = \{(x, y) : (x, y) \in \mathbb{R}^2, ax + by = c\}$$

- Simply put:  $y = \frac{c - ax}{b}$

- + other arithmetic expressions



# Basic concepts – line segments

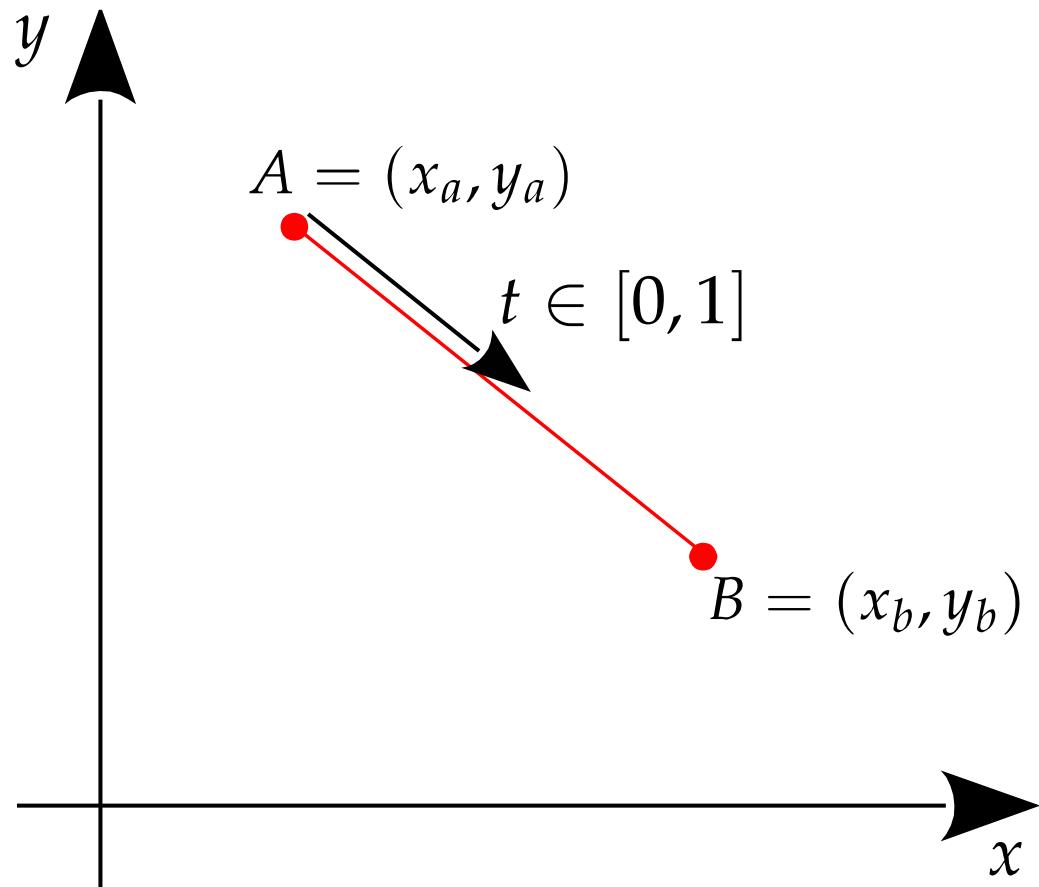
- A line can be parametrized by  $t \in [0,1]$  :

$$\Delta y = y_2 - y_1$$

$$\Delta x = x_2 - x_1$$

$$y = y_1 + t\Delta y$$

$$x = x_1 + t\Delta x$$

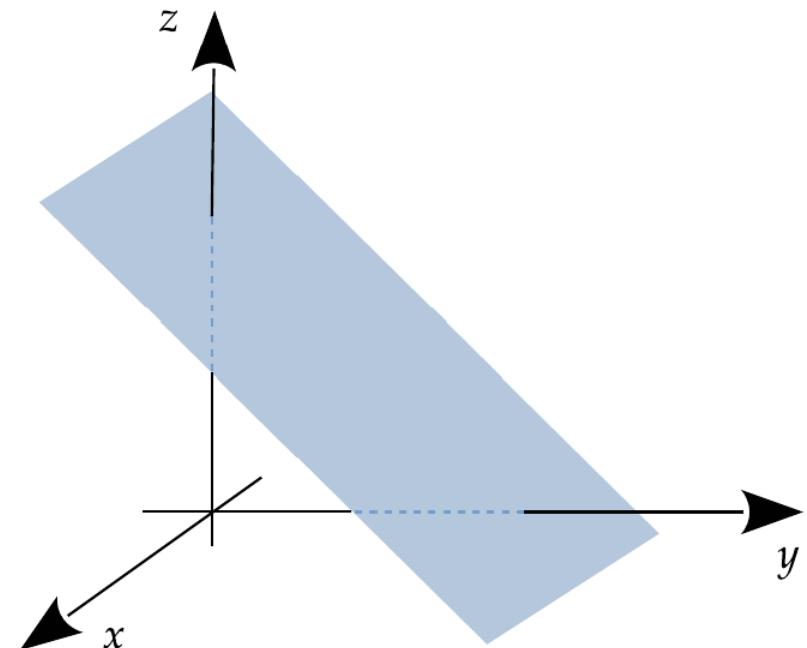


# Basic concepts – plane

- Definition of a plane in 3D:

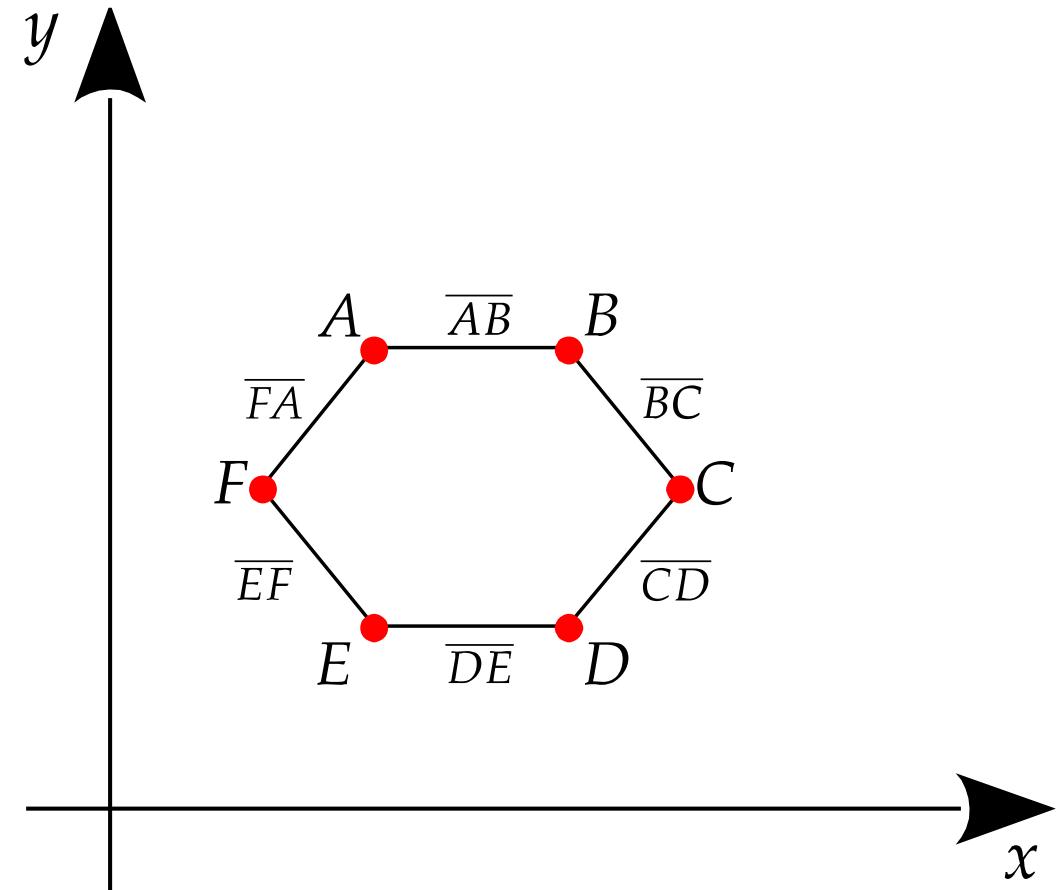
$$P = \{(x, y, z) : (x, y, z) \in \mathbb{R}^3, ax + by + cz = d\}$$

- Geometric shape is a subset of points on a plane



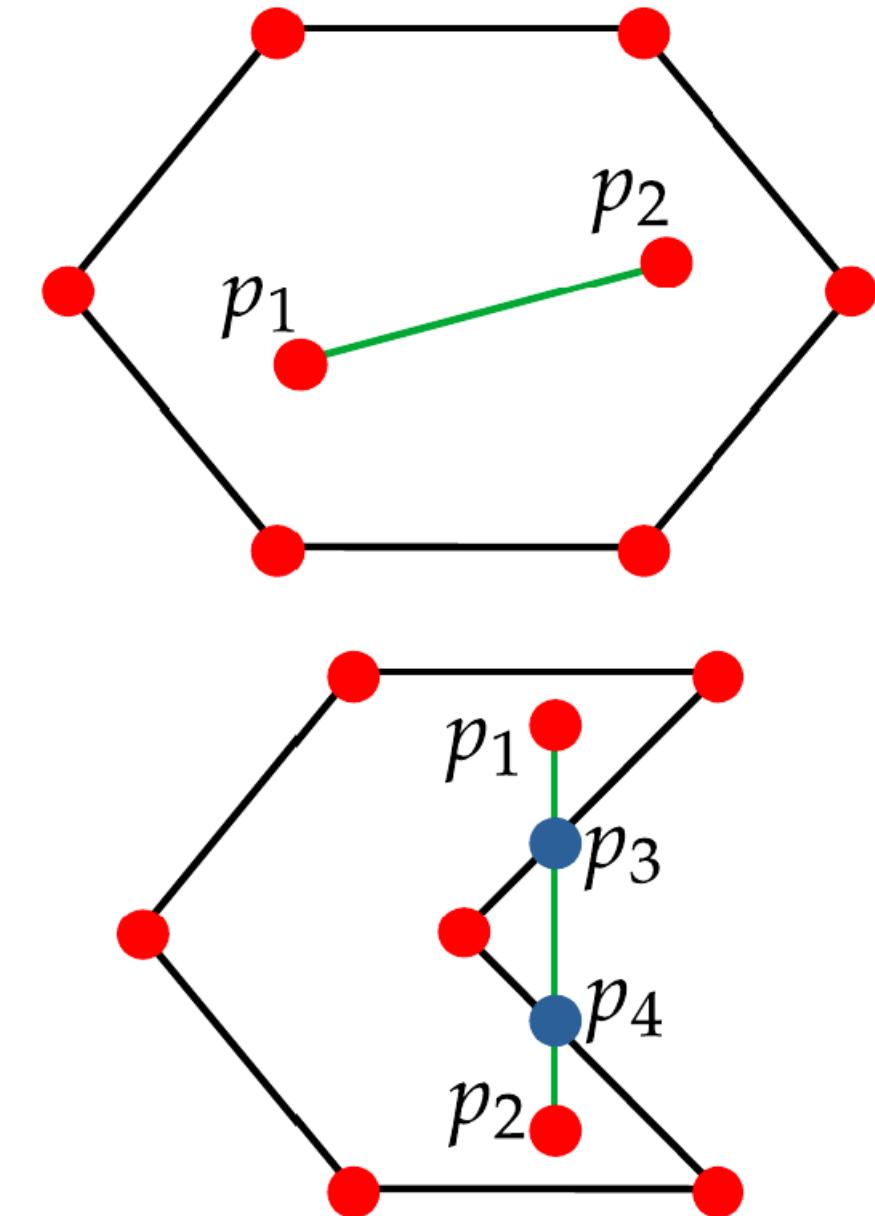
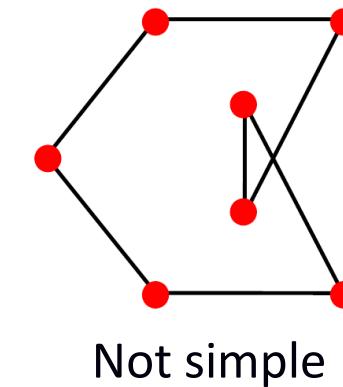
# Basic concepts – polygon (1)

- What's a polygon?
  - Bounded by a subset of points on a plane
  - Bounded by line segments
- Body – internal points
- Chain or edge – line segments



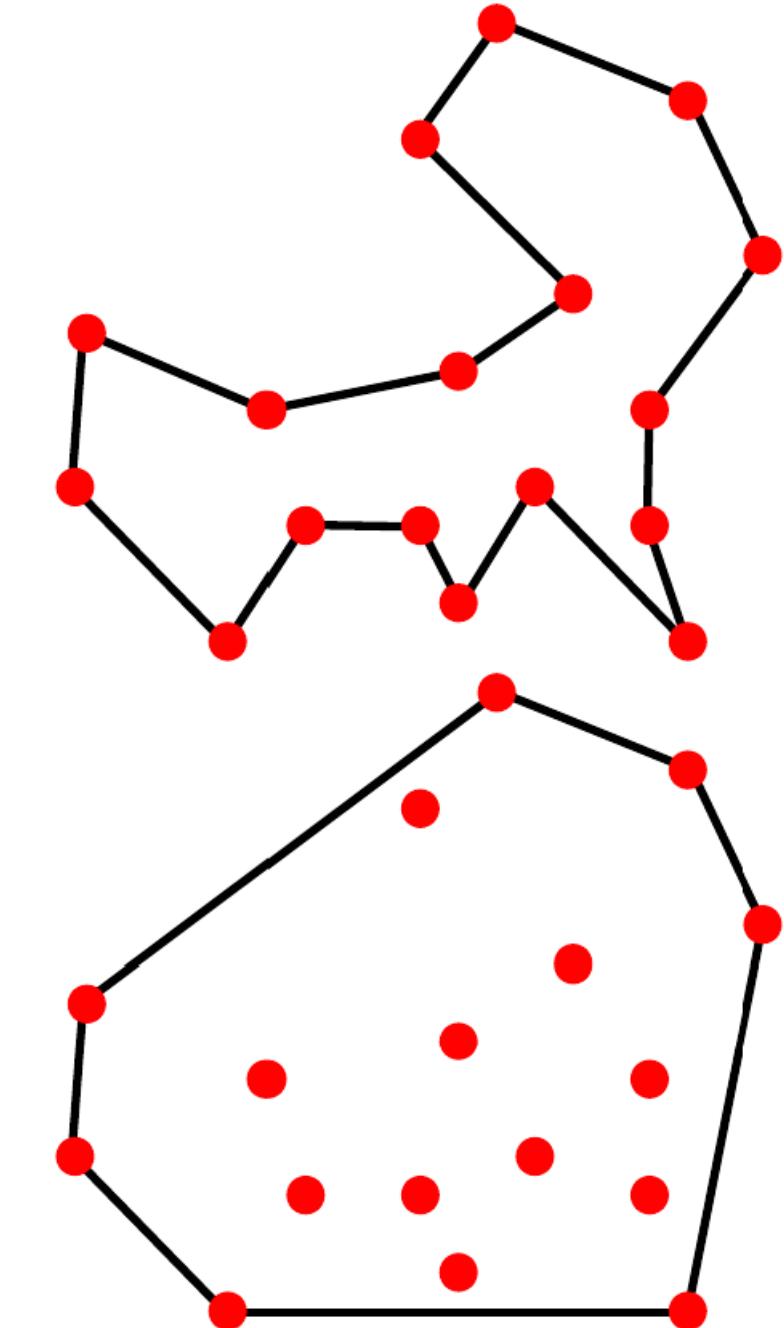
# Basic concepts – polygon (2)

- Convex
  - All line segments of all  $p_1, p_2 \in \mathcal{P}$  are inside the polygon
  - $p_1, p_2 \in \mathcal{P}: \overline{p_1 p_2} \in \mathcal{P}$
- Simple
  - Their edges don't cross
- Regular
  - Equiangular
  - Triangle, square, pentagon...



# Convex hull I (1)

- A set of points  $P_p = \{p_1, p_2, \dots, p_n\}$  in  $\mathbb{R}^2$
- **Task:** Find a subset of  $P_p$  such that they make the edge of a polygon  $\mathcal{P}$  that encompasses all points from  $P_p$   
$$\nexists p \in P_p : p \notin \mathcal{P}$$
- How do we get a CH?
- How do we check if we have it?



# Convex hull I (2) – how to check?

- Pair of points on the hull:

$$\overrightarrow{AB} = [x_b - x_a \quad y_b - y_a]$$

- Set of equations:

$$\overrightarrow{AB} \cdot \vec{n}^T = 0$$

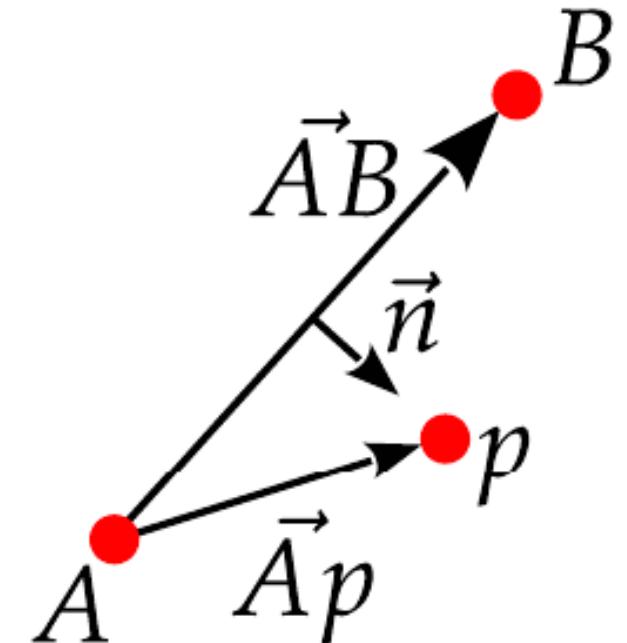
$$D = \vec{n} \cdot \overrightarrow{Ap}^T$$

- Solved to:

$$D = [y_b - y_a \quad x_a - x_b] \begin{bmatrix} x - x_a \\ y - y_a \end{bmatrix}$$

$$D = (x - x_a)(y_b - y_a) - (y - y_a)(x_b - x_a)$$

- $D > 0 \rightarrow p$  is to the right of  $\overrightarrow{AB}$



# Convex hull I (3)

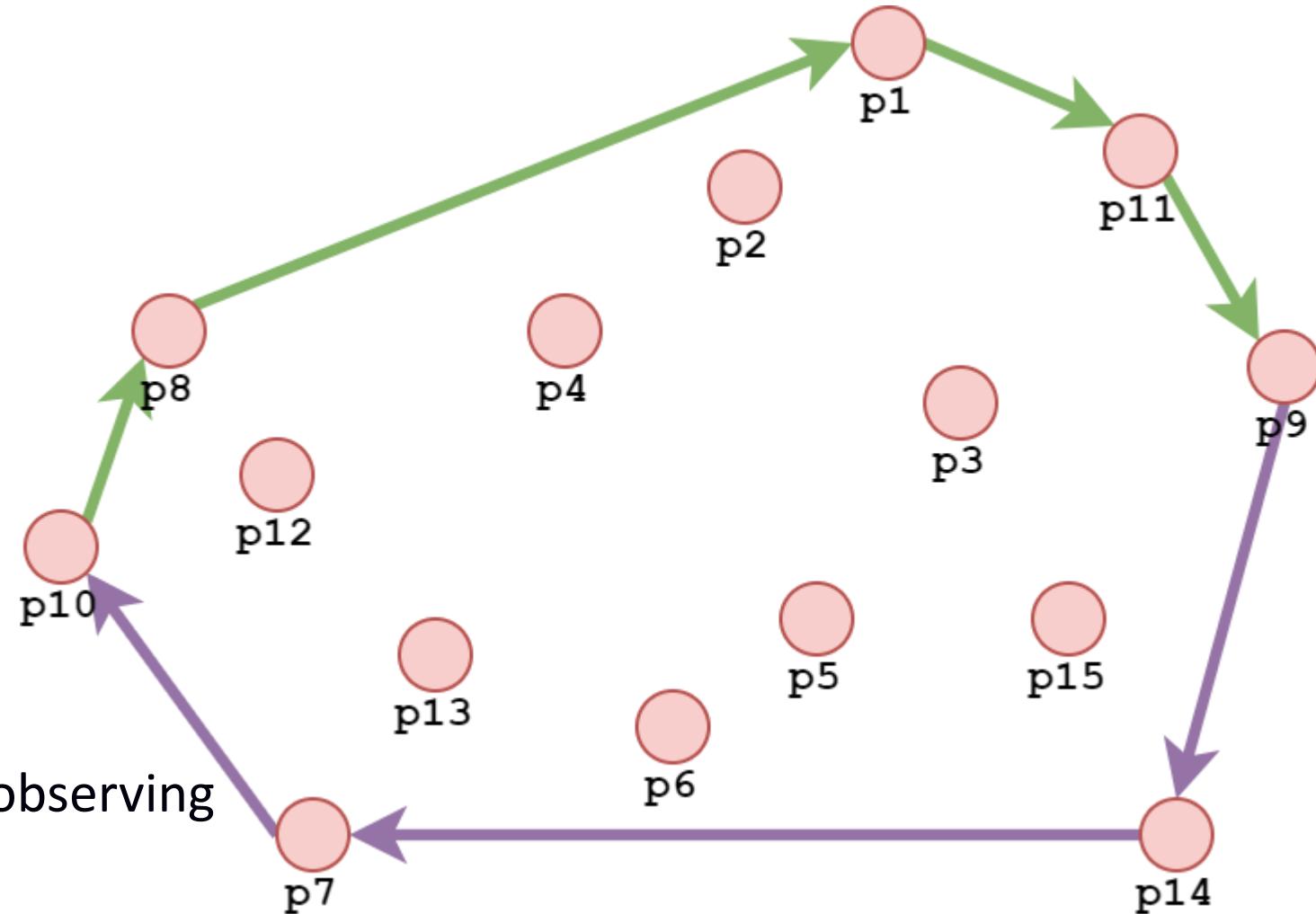
- Go through all pairs of points and find those that have nothing on the left
  - They make the lines of the convex hull

```
for each pair of points (p_i, p_j) in P_p:  
    convex_hull = {}  
    for each point p_k in (P_p \ {p_i, p_j}):  
        if p_k left_of line(p_i, p_j):  
            break #inner loop  
        convex_hull.add(line(p_i, p_j))
```

- $O(n^3)$ 
  - Can we make this faster?

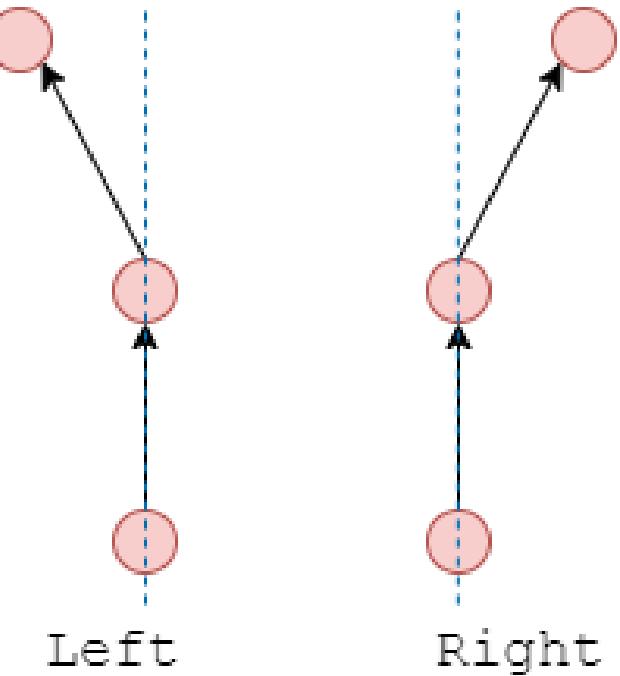
# Convex hull II (1)

- **Strategy:** Split the hull into lower and upper “hull”
- Horizontally sort  $P_p$ 
  - Get  $h(P_p)$
  - $h(P_p) = \{p_{10}, \dots, p_9\}$
- Do inverted passes
  - Avoid left inclination when observing the last three points



# Convex hull II (2)

- Both hulls must have the same incline
  - E.g. **right incline**
  - We need at least 3 points to determine an incline



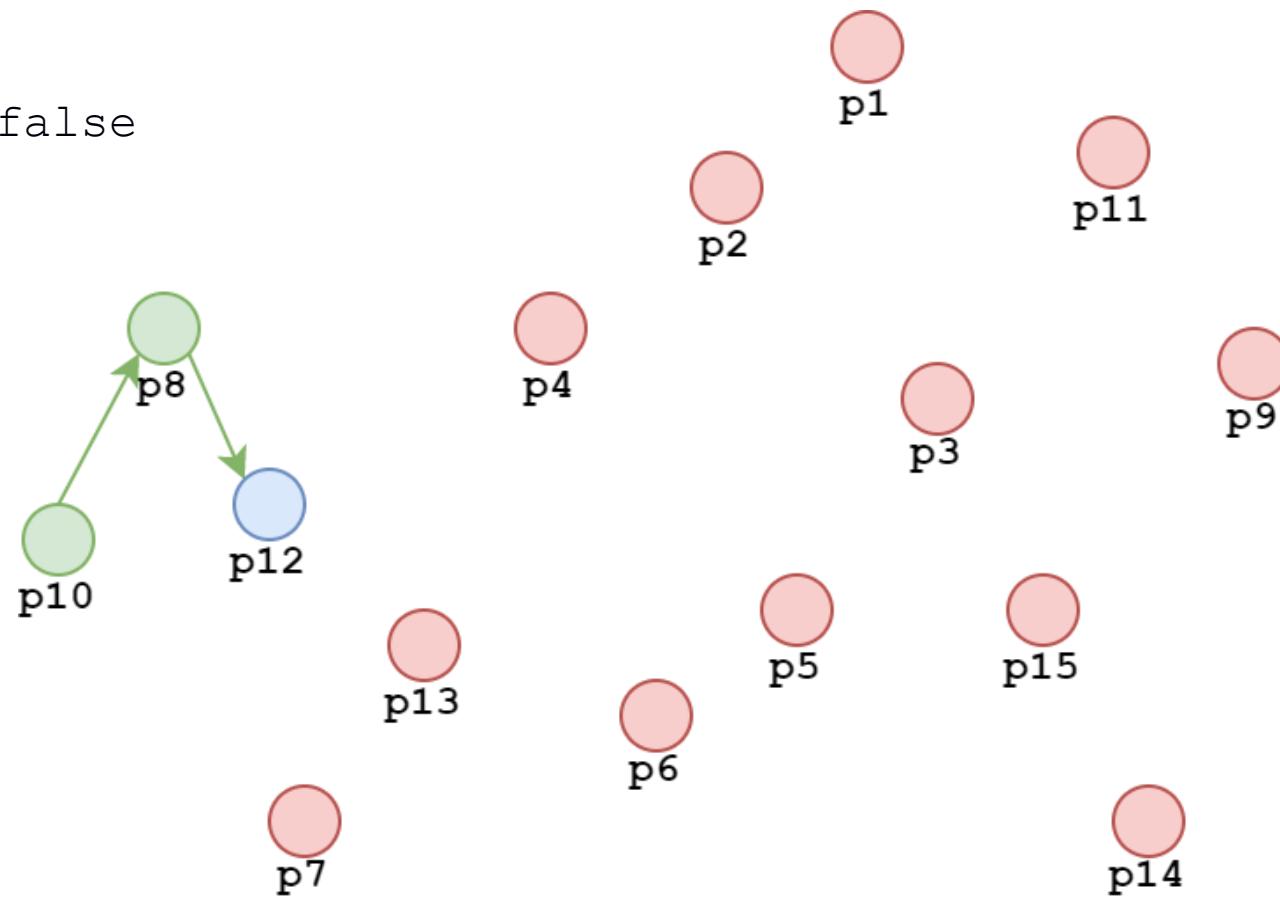
# Convex hull II (3)

```
fun create_convex_hull(points):
    sorted_points = sort(points)
    n = sorted_points.count()
    upper_hull = {sorted_points[0], sorted_points[1]}
    for i in [2 .. n-1]:
        upper_hull.add(sorted_points[i])
        while upper_hull.count() >= 3 and are_inclined_left(p[i-2], p[i-1], p[i]):
            upper_hull.remove(i-1)
    lower_hull = {sorted_points[n-1], sorted[n-2]}
    for i in [n-3 .. 0]:
        lower_hull.add(sorted_points[i])
        while lower_hull.count() >= 3 and are_inclined_left(p[i], p[i-1], p[i-2]):
            lower_hull.remove(i+1)
    upper_hull + lower_hull
```

- Sort:  $O(n \log n)$
- Passthrough:  $O(n)$
- Completely:  $O(n \log n + n) = O(n \log n)$

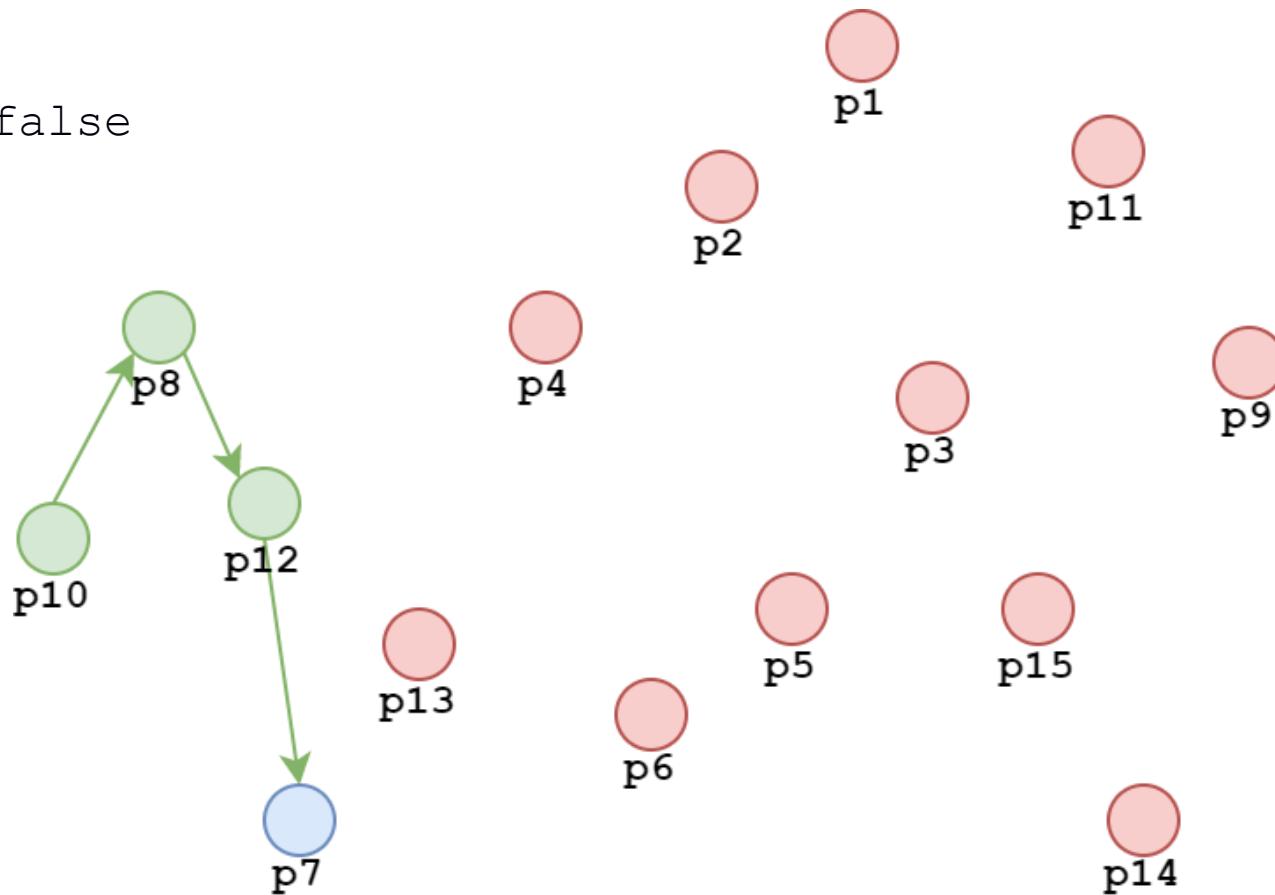
# Convex hull II (4)

```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}  
upper_hull = {p10, p8, p12}  
i = 2  
is_left_incline = false
```



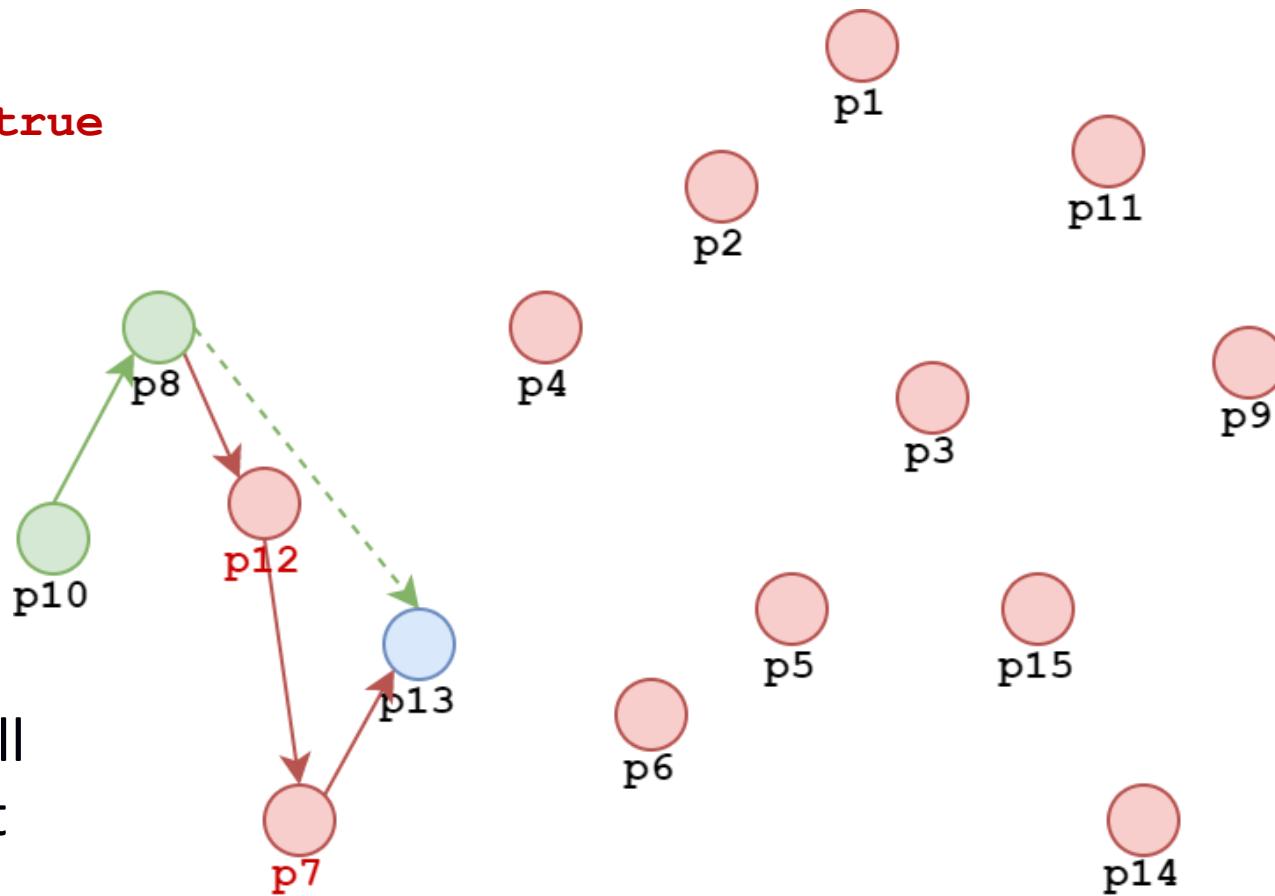
# Convex hull II (5)

```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}  
upper_hull = {p10, p8, p12, p7}  
i = 3  
is_left_incline = false
```



# Convex hull II (6)

```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}  
upper_hull = {p10, p8, p12, p7, p13}  
i = 4  
is_left_incline = true
```



# Convex hull II (7)

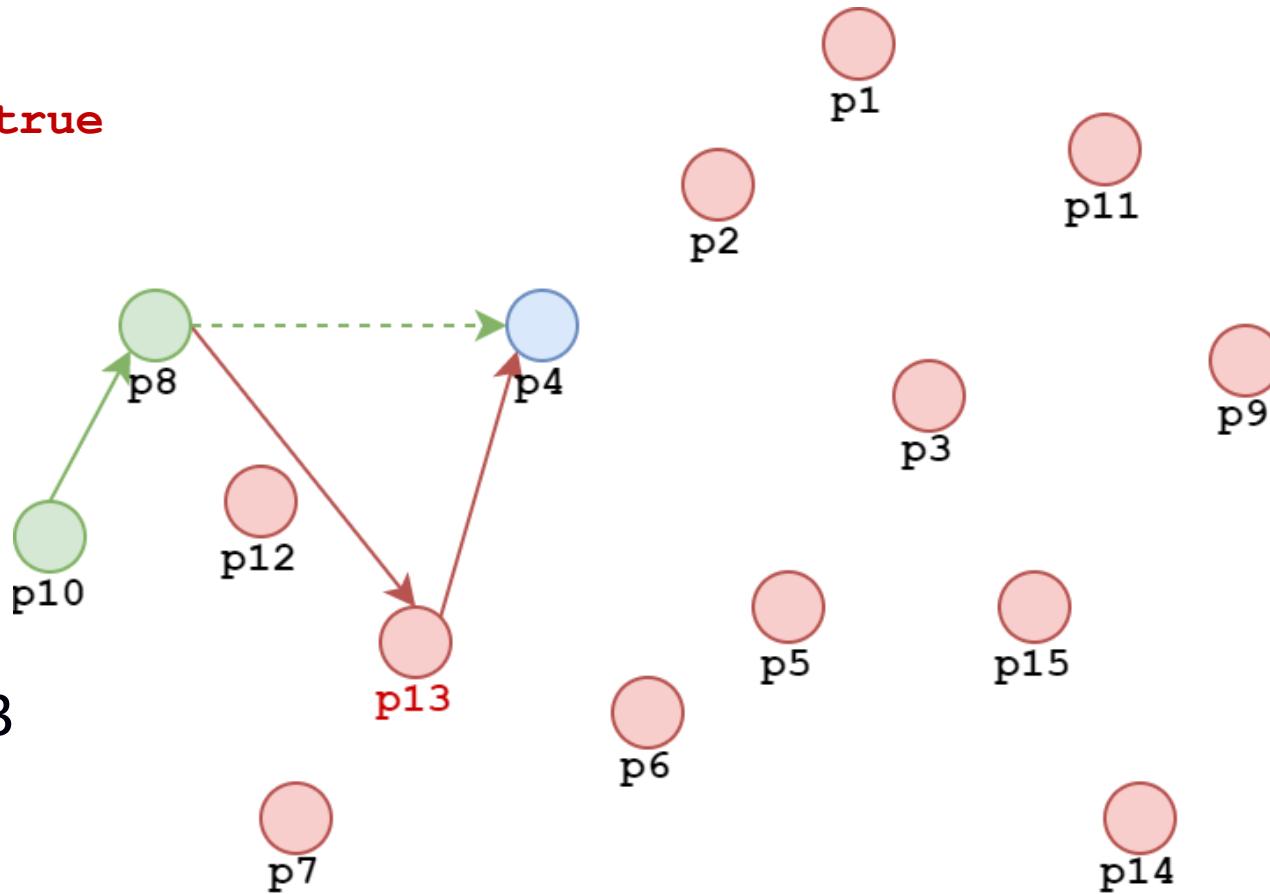
```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}
```

```
upper_hull = {p10, p8, p13, p4}
```

```
i = 5
```

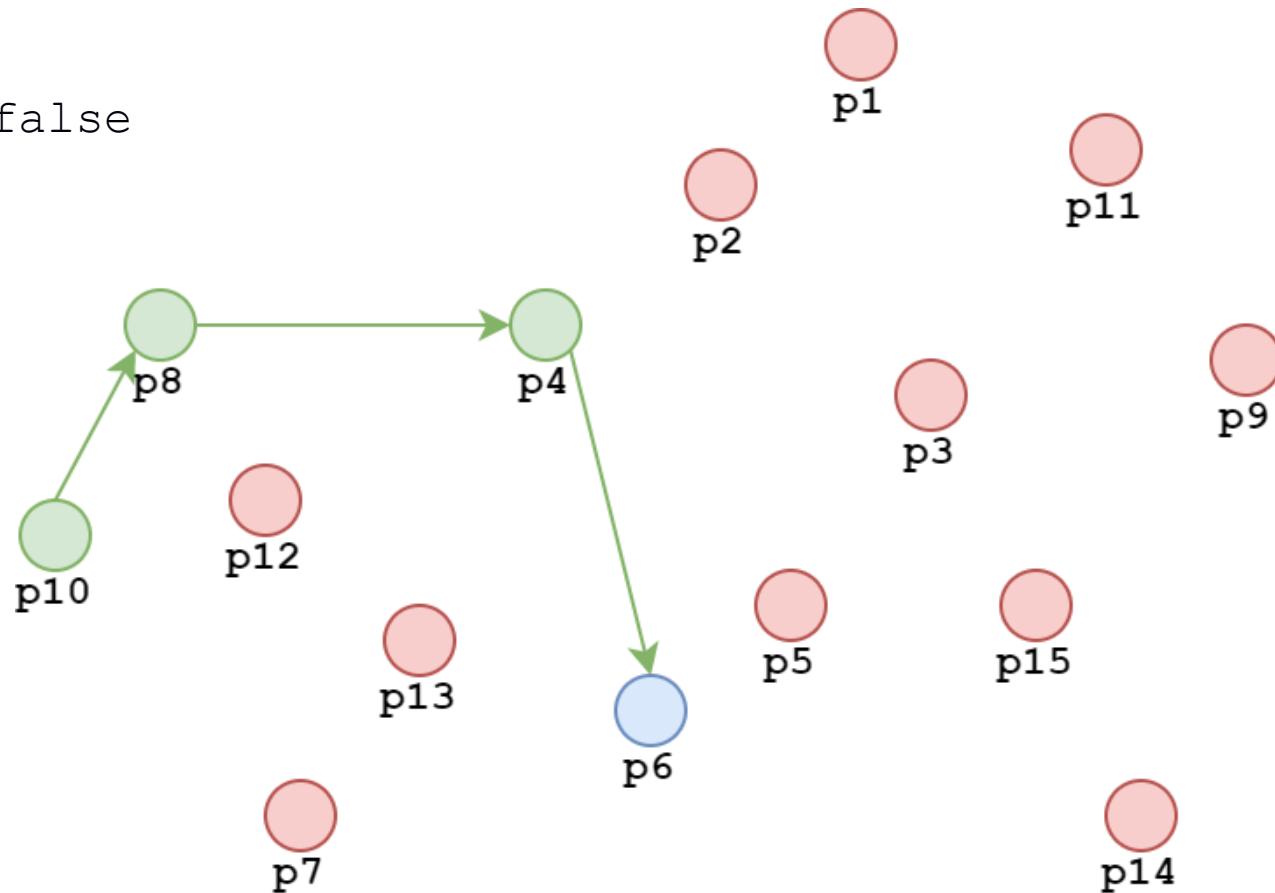
```
is_left_incline = true
```

Second last is now 13



# Convex hull II (8)

```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}  
upper_hull = {p10, p8, p4, p6}  
i = 6  
is_left_incline = false
```



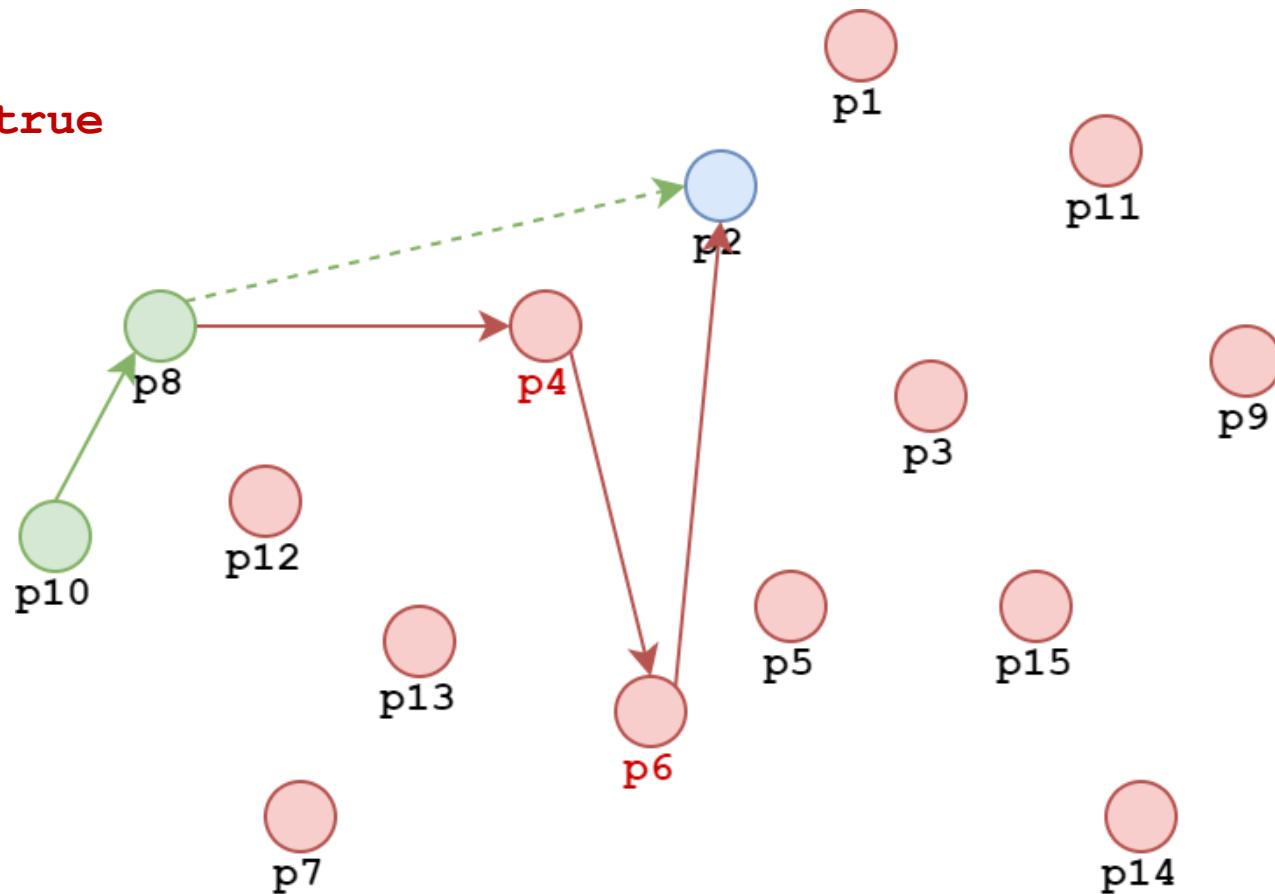
# Convex hull II (9)

```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}
```

```
upper_hull = {p10, p8, p4, p6, p2}
```

```
i = 7
```

```
is_left_incline = true
```



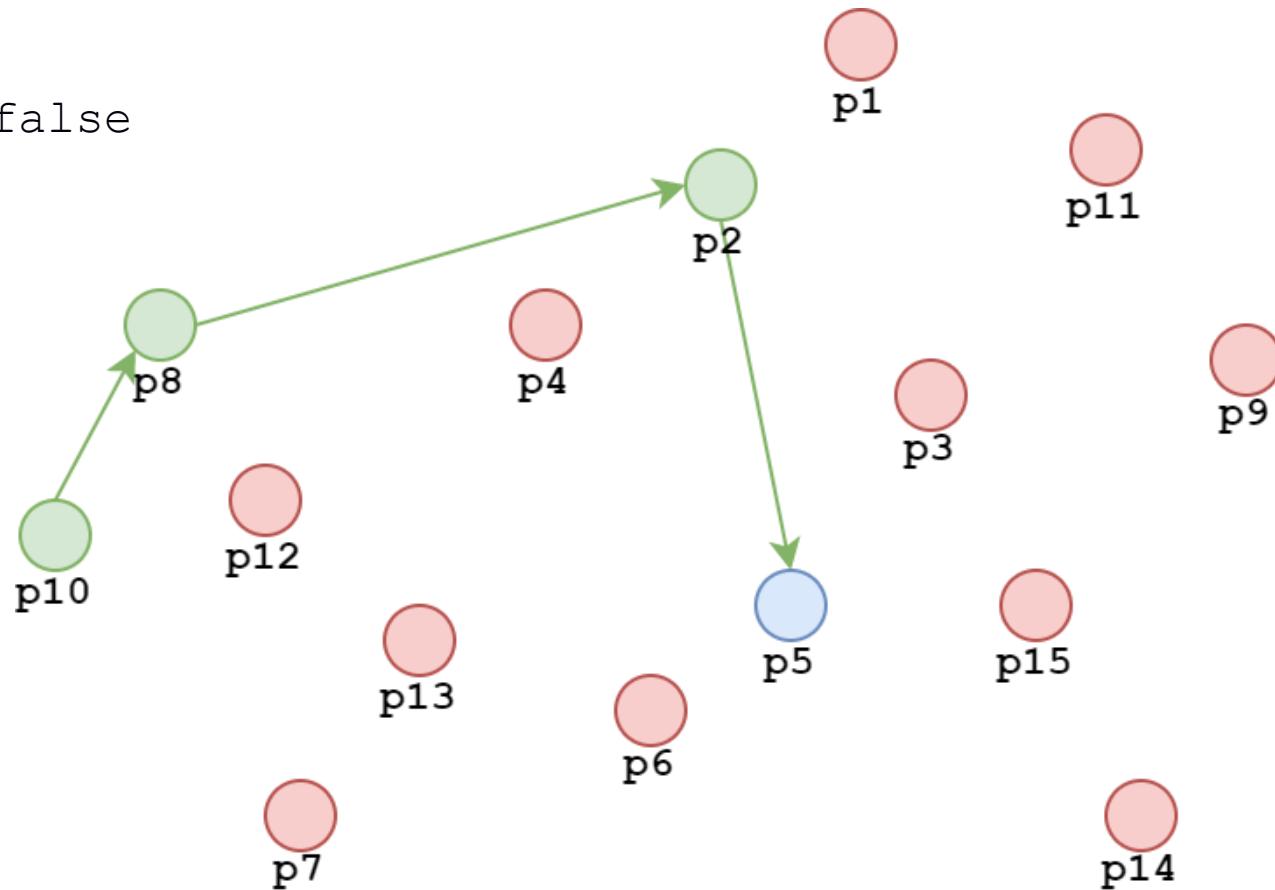
# Convex hull II (10)

```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p1, p3, p15, p11, p14, p9}
```

```
upper_hull = {p10, p8, p2, p5}
```

```
i = 8
```

```
is_left_incline = false
```



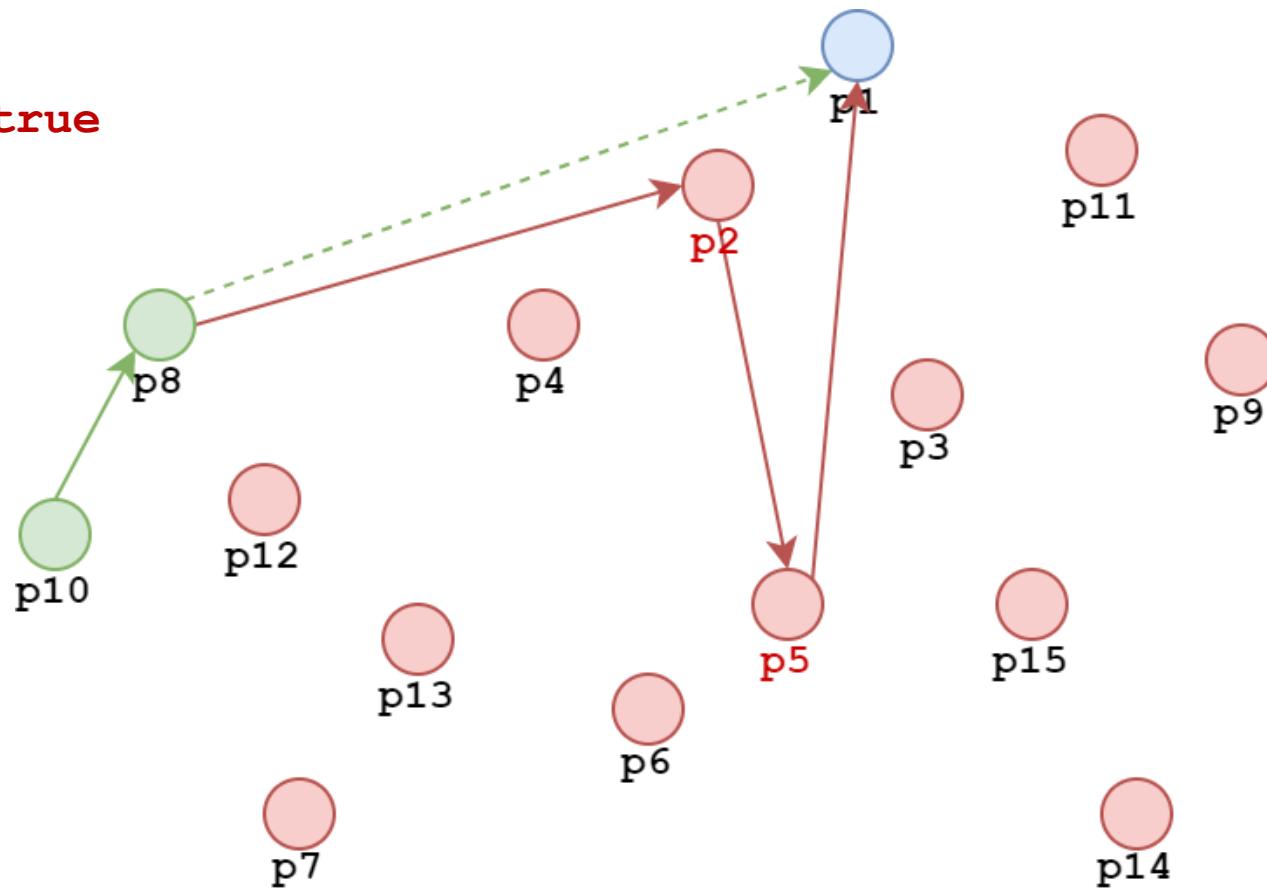
# Convex hull II (11)

```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}
```

```
upper_hull = {p10, p8, p2, p5, p1}
```

```
i = 9
```

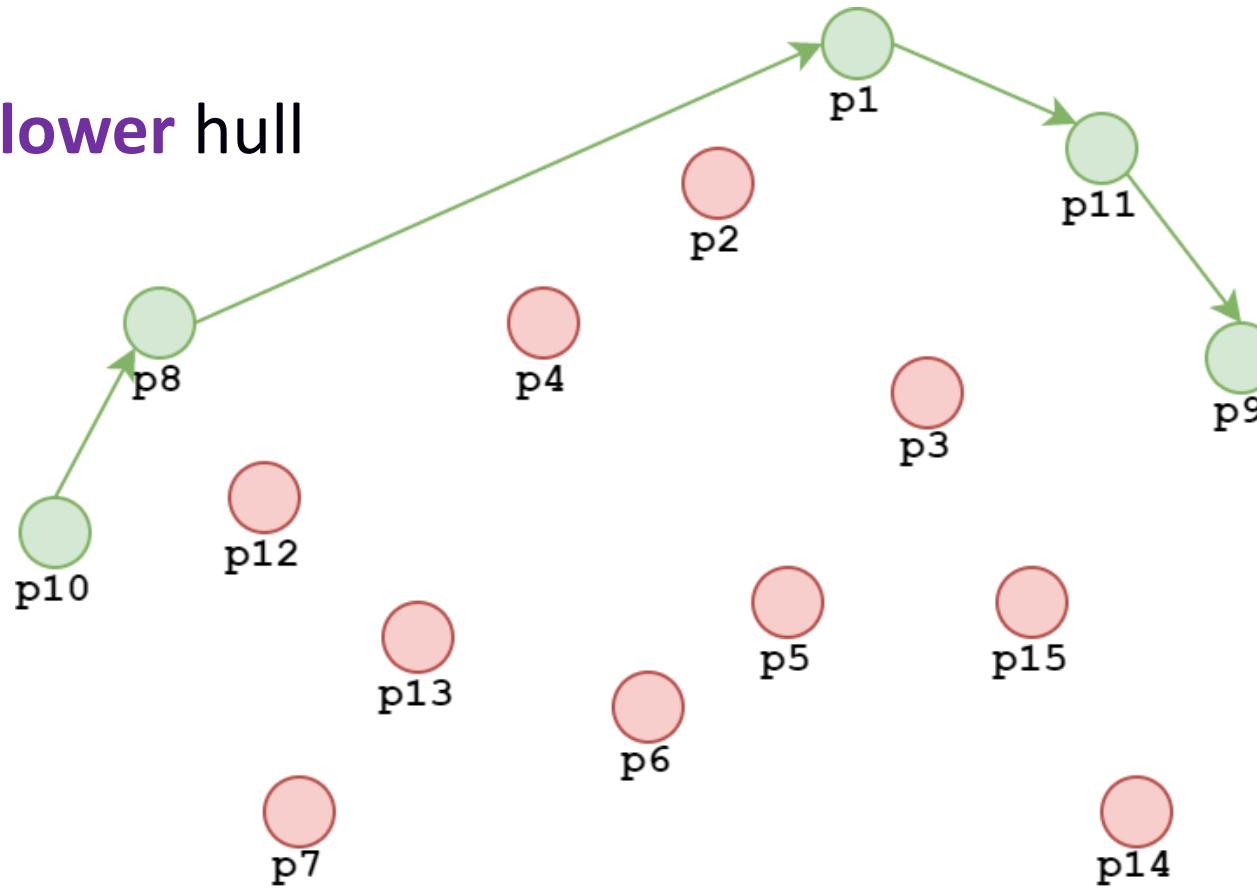
```
is_left_incline = true
```



# Convex hull II (11)

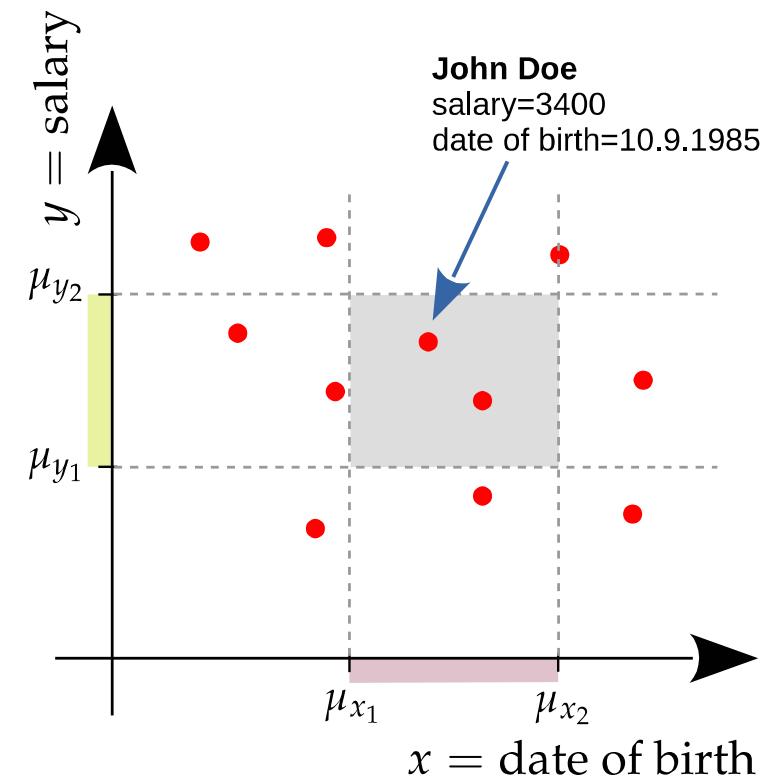
```
sorted_points = {p10, p8, p12, p7, p13, p4, p6, p2, p5, p1, p3, p15, p11, p14, p9}  
upper_hull = {p10, p8, p1, p11, p9}
```

- And then the **lower** hull



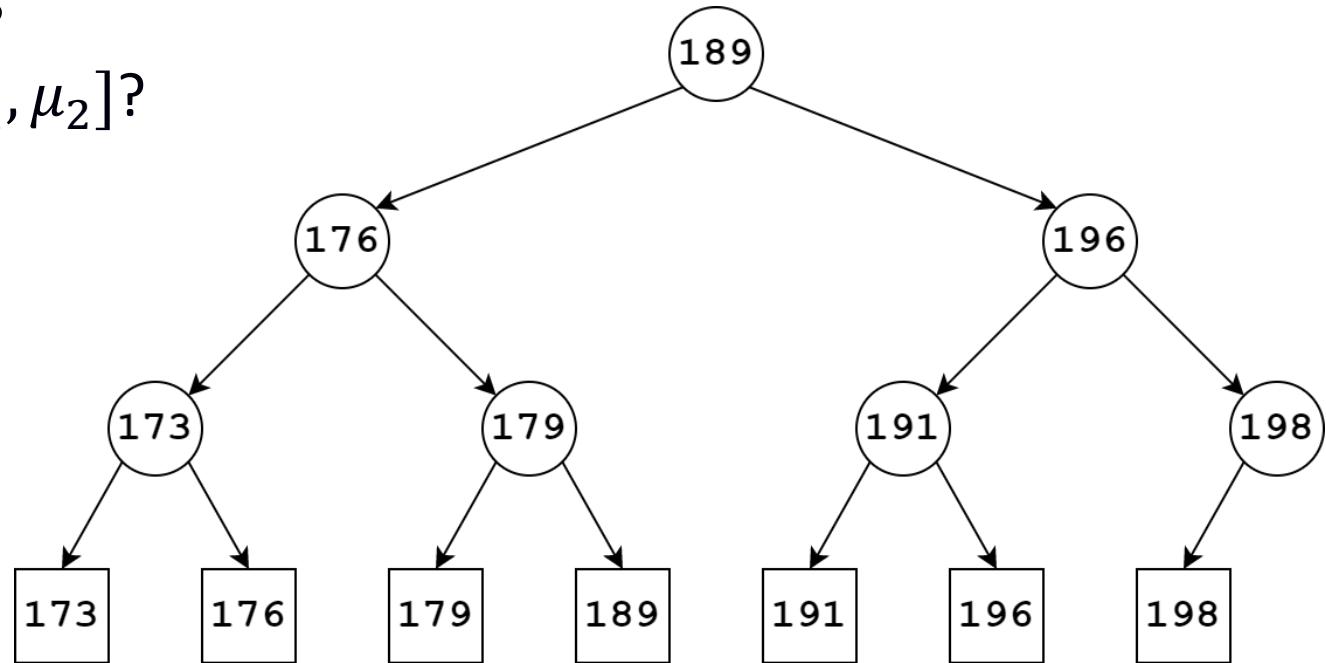
# Geometric search

- N variables
  - N dimensional search space
- Height, weight, salary...
- Databases
  - B+-trees?



# 1D range search – Binary range trees

- Use-case: 1D range queries
  - What keys exist in range  $[\mu_1, \mu_2]$ ?
- Children as leaves
- Different from B+ trees?



# Binary range trees – search

- Find the splitting node  $n_{split}$  (ordinary traversal)
  - The first node within the range  $[\mu_1, \mu_2]$
- Initiate left and right search
  - Skip redundant subtrees

## $n_{split}$ left subtree

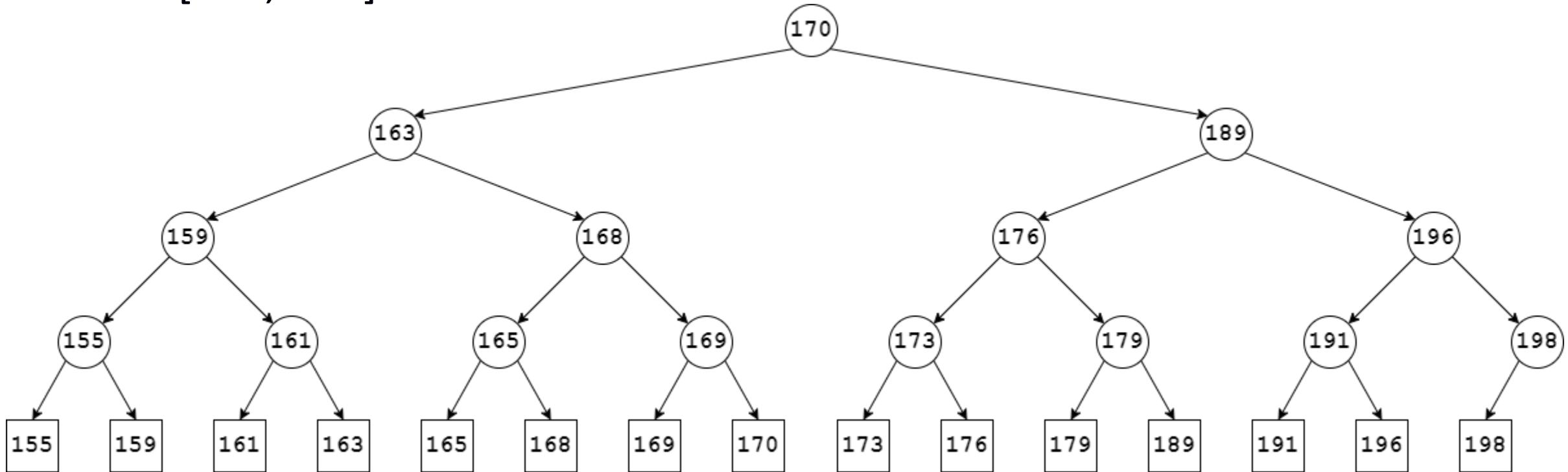
```
if  $\mu_1 \leq n$  :  
    take right subtree  
    traverse to ChL(n)  
  
if  $\mu_1 > n$  :  
    skip  
    traverse to ChR(n)
```

## $n_{split}$ right subtree

```
if  $\mu_2 \geq n$  :  
    take left subtree  
    traverse to ChR(n)  
  
if  $\mu_2 < n$  :  
    skip  
    traverse to ChL(n)
```

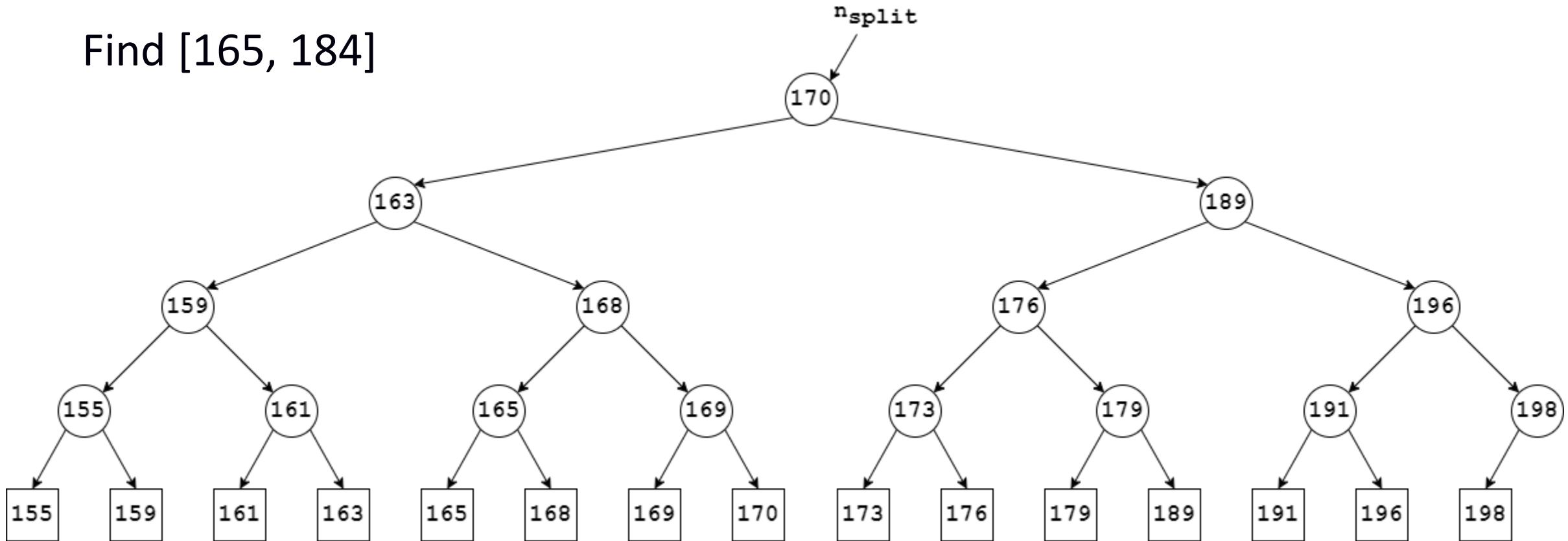
# Binary range trees – search example (1)

Find [165, 184]



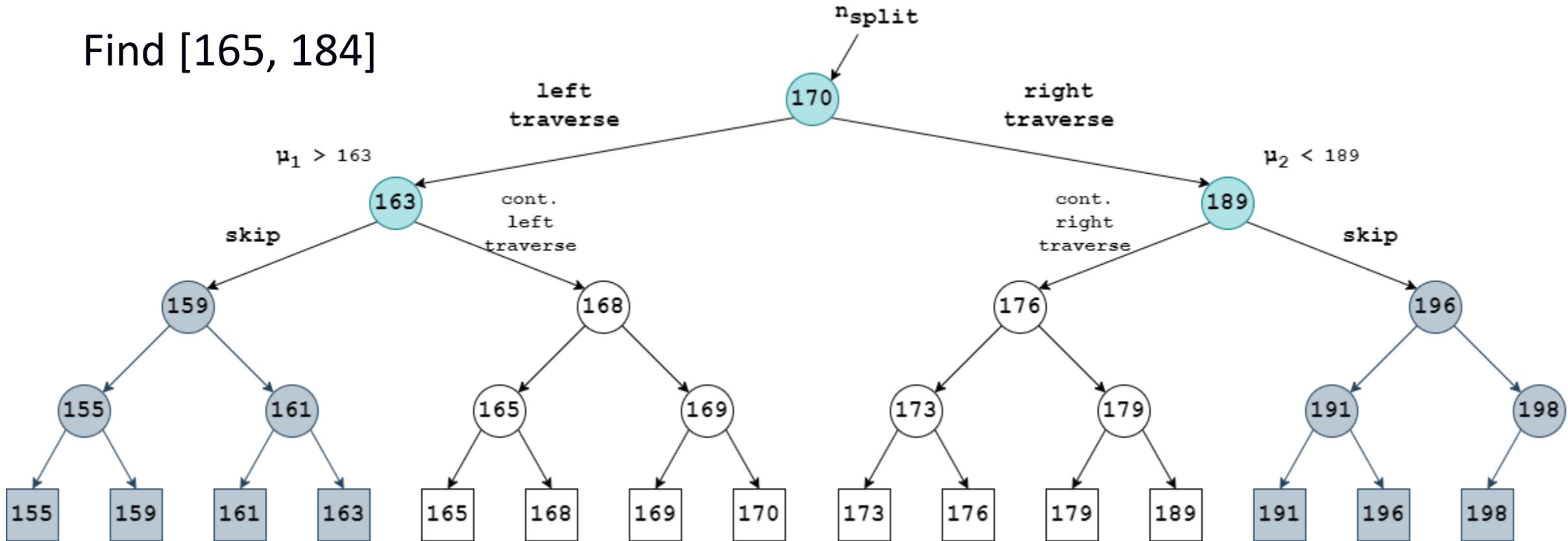
# Binary range trees – search example (2)

Find [165, 184]



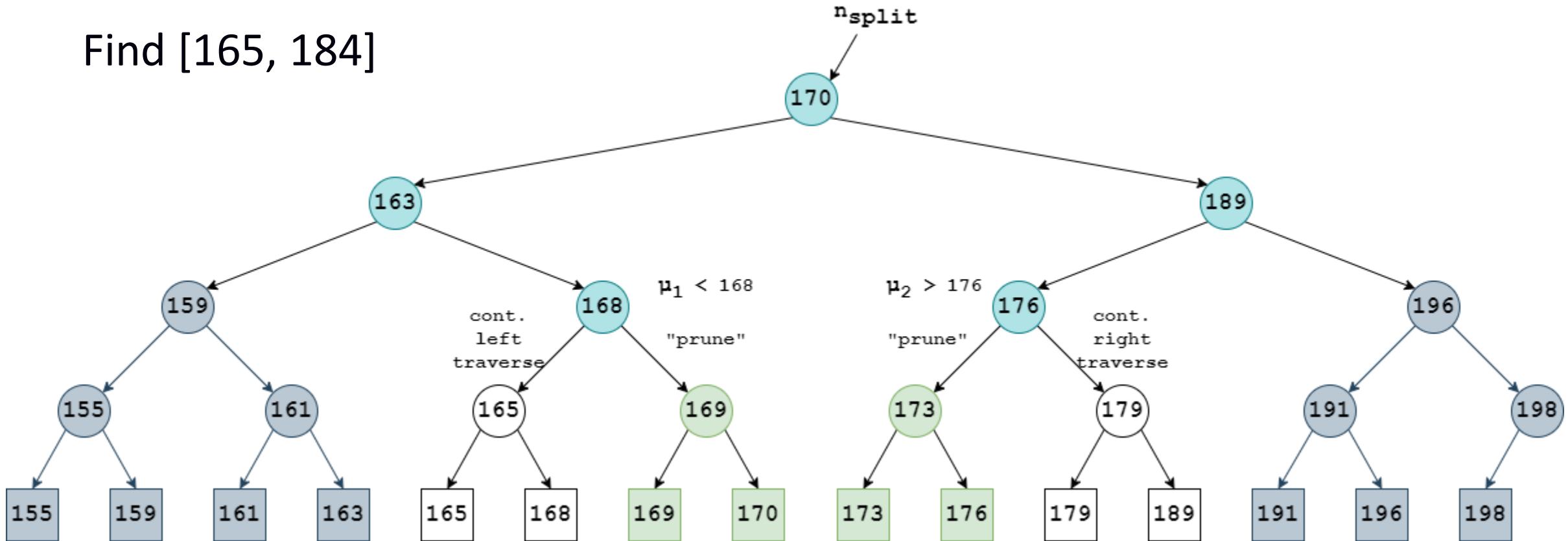
# Binary range trees – search example (3)

Find [165, 184]



# Binary range trees – search example (4)

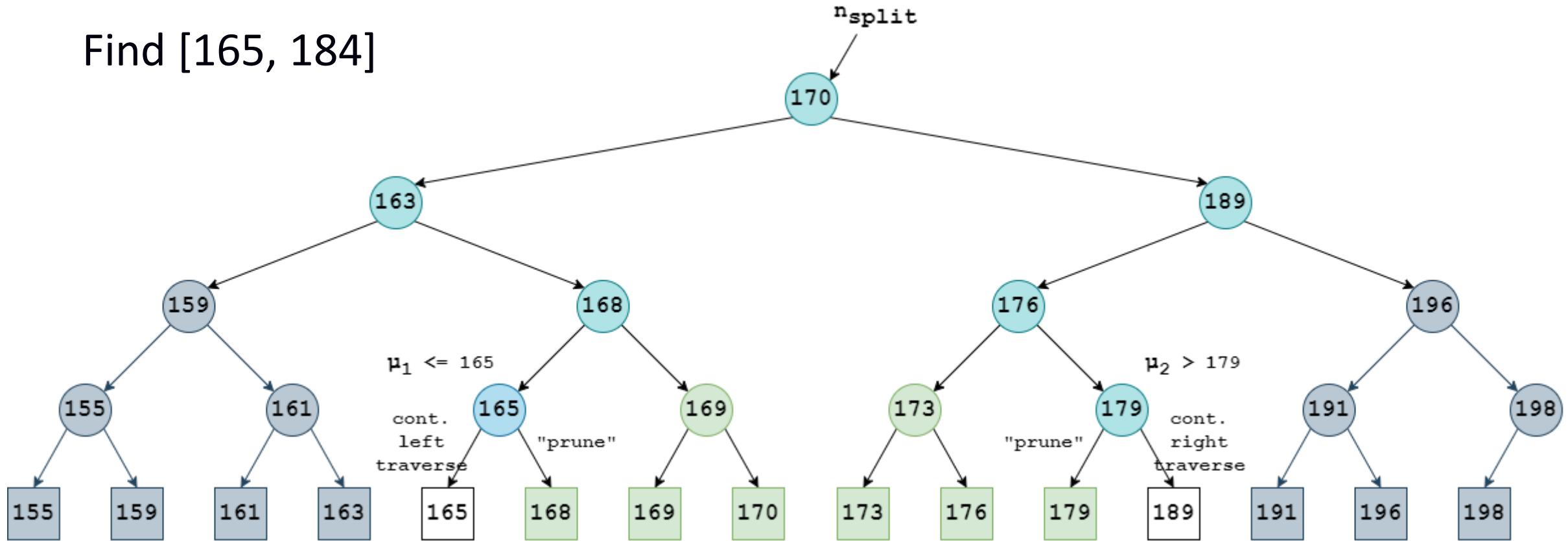
# Find [165, 184]



\*“prune” as in “just take everything”

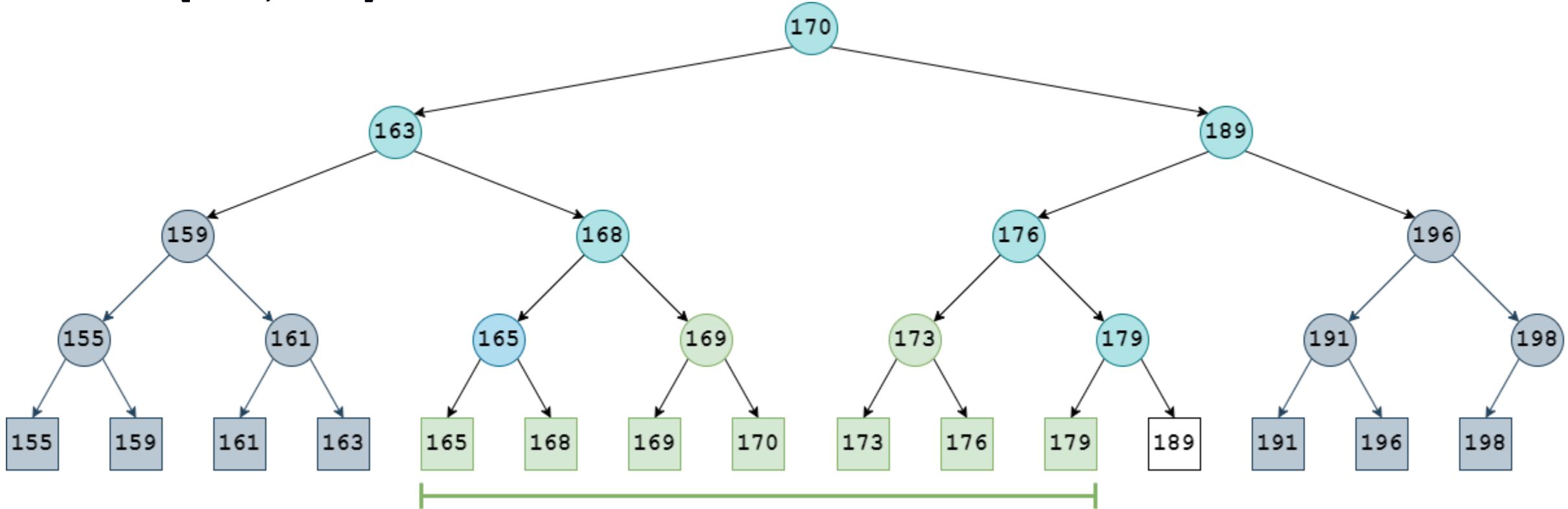
# Binary range trees – search example (5)

Find [165, 184]



# Binary range trees – search example (6)

Find [165, 184]



Leaves reached – DONE!

# Binary range trees – search

- 2 parts (functions):
  - Find splitting node (root in example)
  - Right + Left traversal
- Complexity:  $O(k + \log n)$
- $k$  nodes pruned
- $\log n$  binary traversal search

# Binary range trees – creation (1)

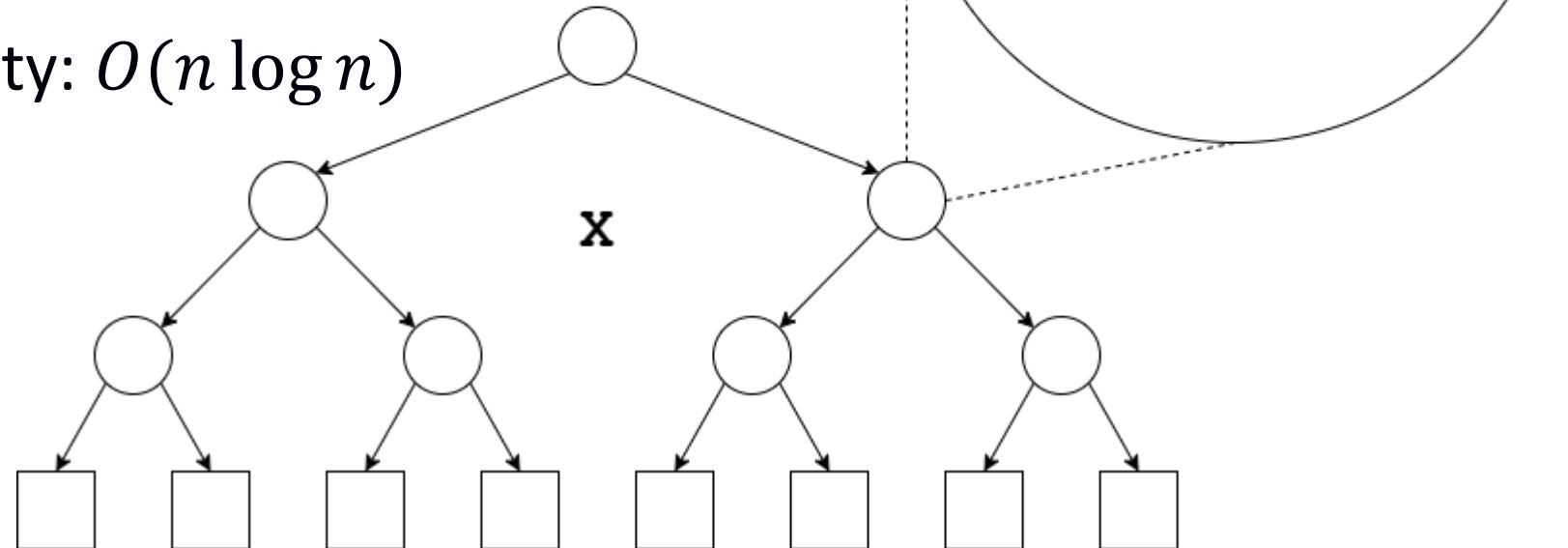
- Just like a sorted binary tree
- But we take medians as root and internal nodes' values
- Take an array and split it over medians
- Complexity:  $O(n \log n)$

# Binary range trees – creation (2)

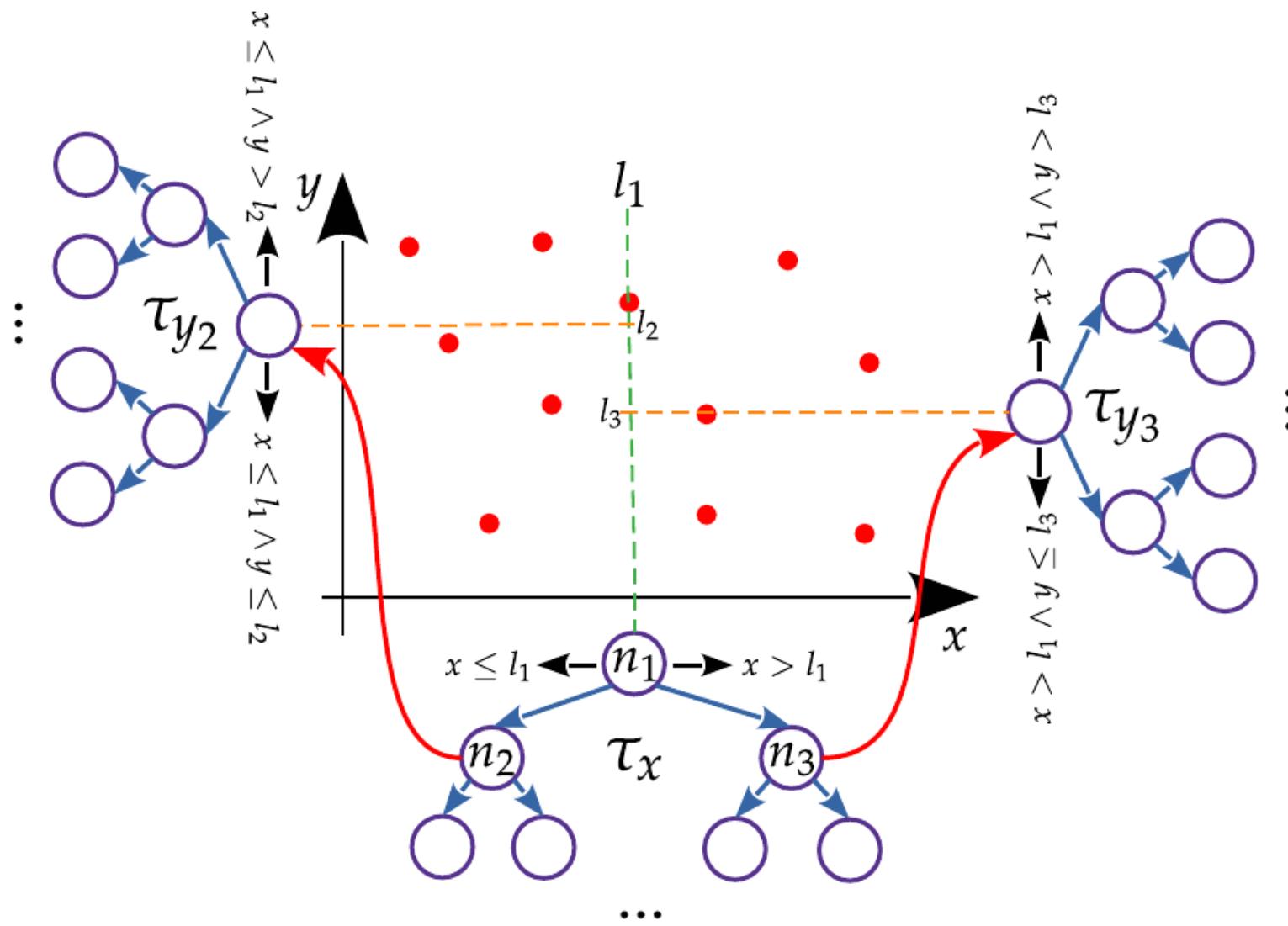
```
fun create_1d_range_tree(points):
    if points.count() == 1:
        points[0] # as only and root node
    else:
        curr_median = median(points)
        points_left = [for point in points if point <= curr_median]
        points_right = [for point in points if point > curr_median]
        left_subtree = create_1d_range_tree(points_left)
        right_subtree = create_1d_range_tree(points_right)
        curr_node = Node(curr_median)
        curr_node.left = left_subtree
        curr_node.right = right_subtree
        curr_node
```

# 2D range search

- 2D range tree
- Every node of a tree is ANOTHER TREE
  - Each  $x$  in the  $X$  range has its own  $Y$  range
- Search complexity:  $O(k + \log^2 n)$
- Creation complexity:  $O(n \log n)$



# 2D range search

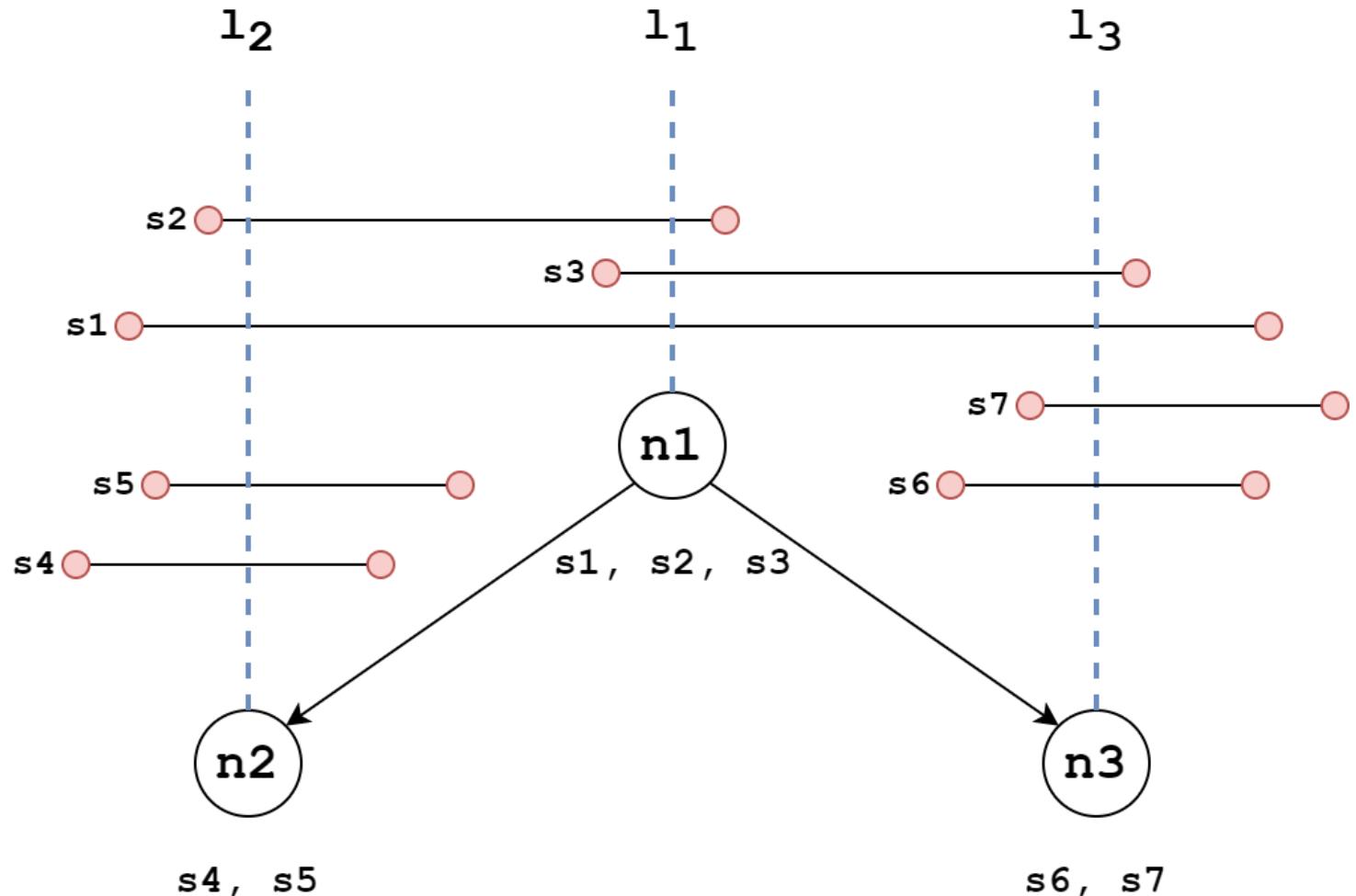


# “1D” interval search

- A set of intervals:  $I = \{s_i = ([x_{i_1}, x_{i_2}], y_i) : x_{i_1}, x_{i_2}, y_i \in \mathbb{R}\}$ 
  - Parallel with x-axis ( $y$  is const. per interval)
  - Not really 1D since, we can take into account the  $y$ -axis
- Each interval has 2 end points:
  - Left:  $(x_{i_1}, y_i)$
  - Right:  $(x_{i_2}, y_i)$
- Interval tree

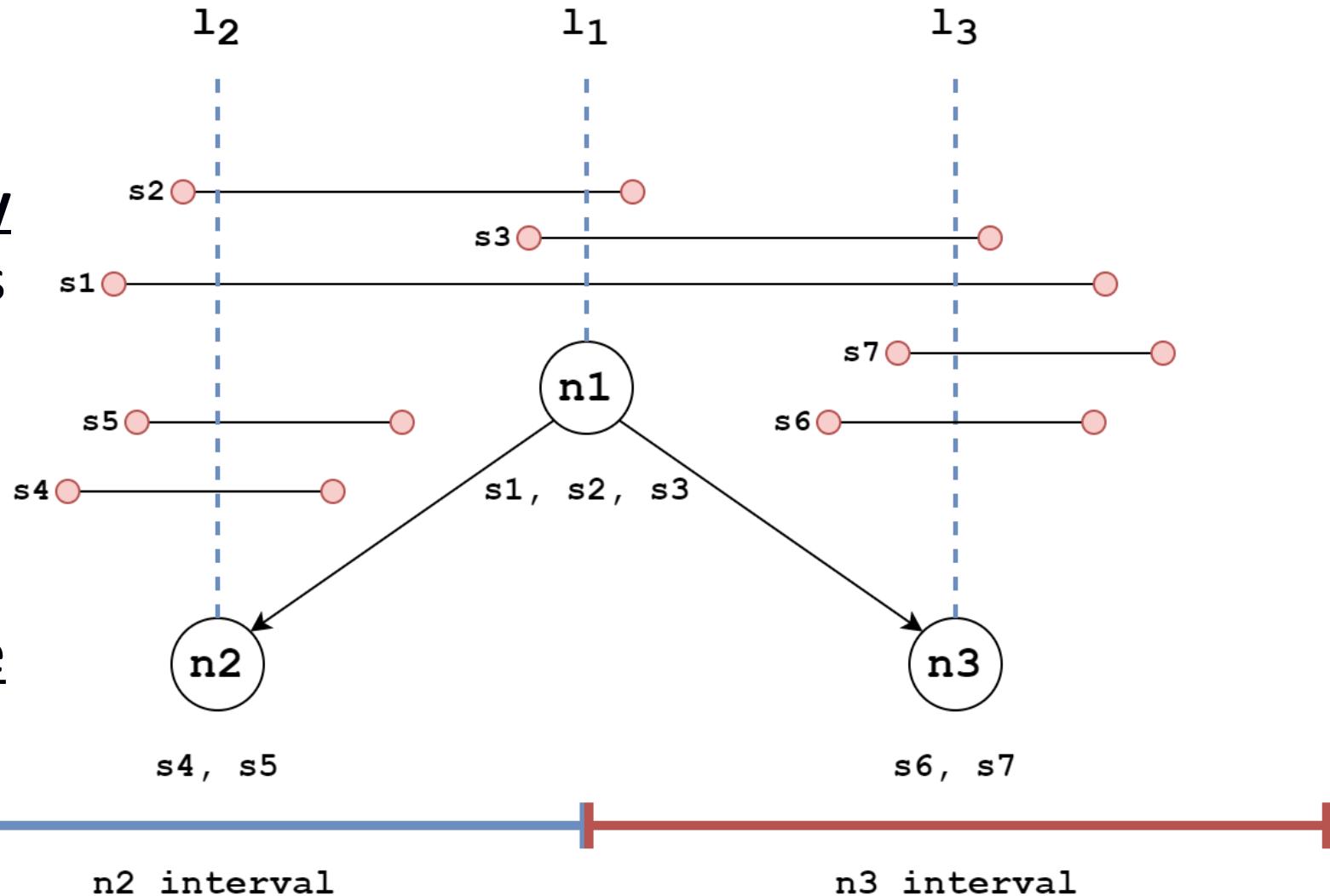
# Interval tree (1)

- Sorted around medians:
  - $l_i = \text{med}(n_i)$
  - defined by the intervals in the node
- Why isn't  $s_3$  in  $n_3$ ?
- Why isn't  $s_2$  in  $n_2$ ?
- We sort by medians, but emulate ranges!



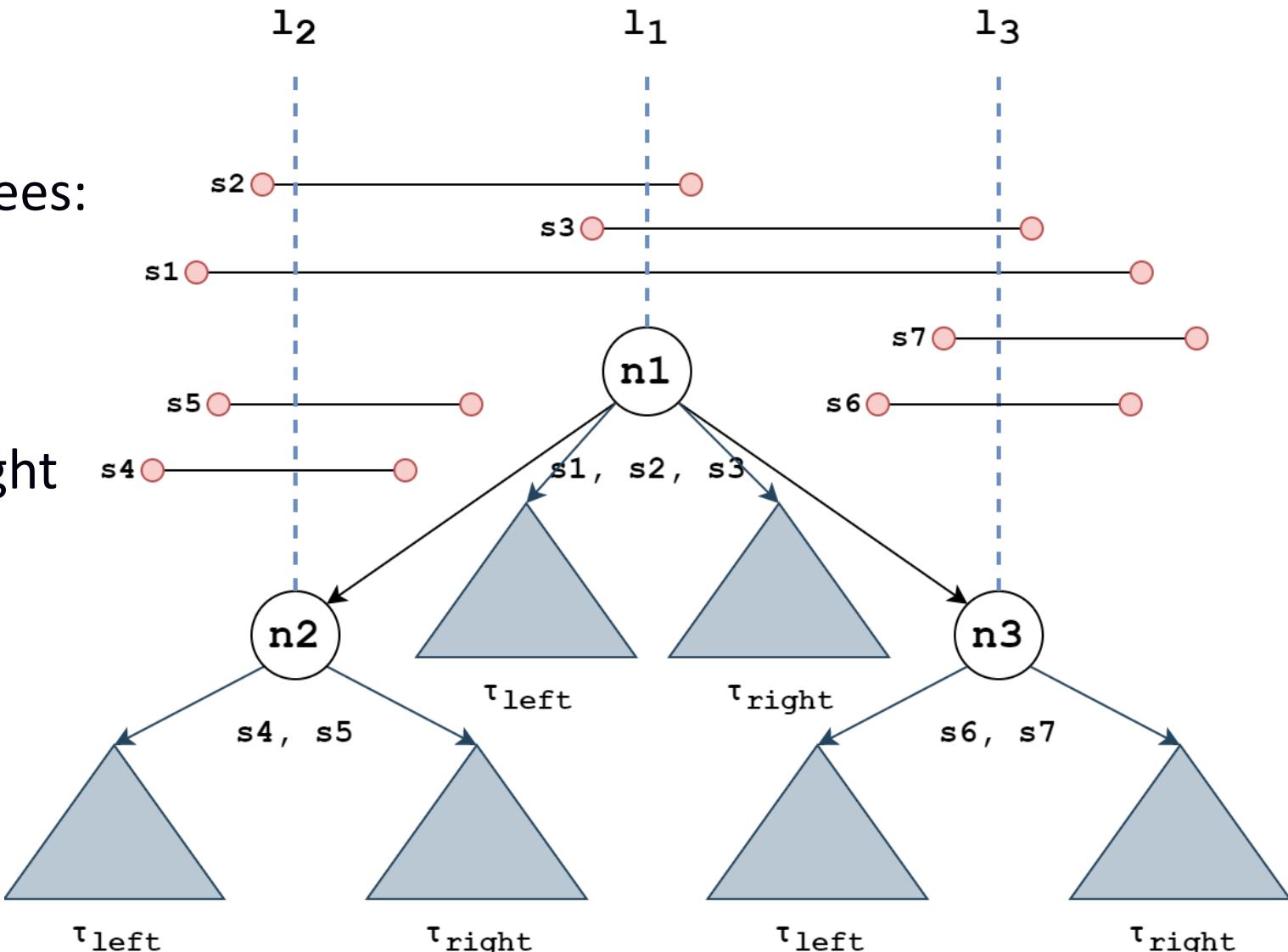
# Interval tree (2)

- $s_1$ ,  $s_2$ , and  $s_3$  don't belong exclusively to the  $n_2$  or  $n_3$  intervals
- They belong exclusively to  $n_1$
- Both endpoints must be in the interval



# Interval tree (3)

- Each node  $n_i$  has 2 range trees:  
 $\tau_{left}$  and  $\tau_{right}$ 
  - Or any other trees
- They contain the left and right endpoints for the intervals
- e.g.  $n_2$  has:
  - $\tau_{left}: s_{4left}, s_{5left}$
  - $\tau_{right}: s_{4right}, s_{5right}$



# Interval tree – creation (1)

```
fun create_interval_tree(intervals):
    if intervals.count() == 0:
        Node() # empty
    else:
        curr_median = median([i.endpoints() for i in intervals])
        intervals_l = [i for i in intervals
                      if i.endp_l < curr_median and i.endp_r < curr_median] # sort
        intervals_r = [i for i in intervals
                      if i.endp_l > curr_median and i.endp_r > curr_median] # sort
        node_intervals = intervals \ (intervals_l + intervals_r) # leftovers
        curr_node = Node(curr_median)
        curr_node.ch_l = create_interval_tree(intervals_l)
        curr_node.ch_r = create_interval_tree(intervals_r)
        if node_intervals.count() != 0:
            endp_l = [i.endp_l for i in node_intervals if i.endp_l <= curr_median]
            endp_r = [i.endp_r for i in node_intervals if i.endp_r > curr_median]
            curr_node.tau_l = create_2d_range_tree(endp_l) # range tree
            curr_node.tau_r = create_2d_range_tree(endp_r) # range tree
        curr_node
```

# Interval tree – creation (2)

- Sorting:  $O(n \log n)$
- 2 range trees per node:  $O((n \log n)(2 \log n))$
- Completely:  $O(n \log^2 n)$

# Interval tree – search (1)

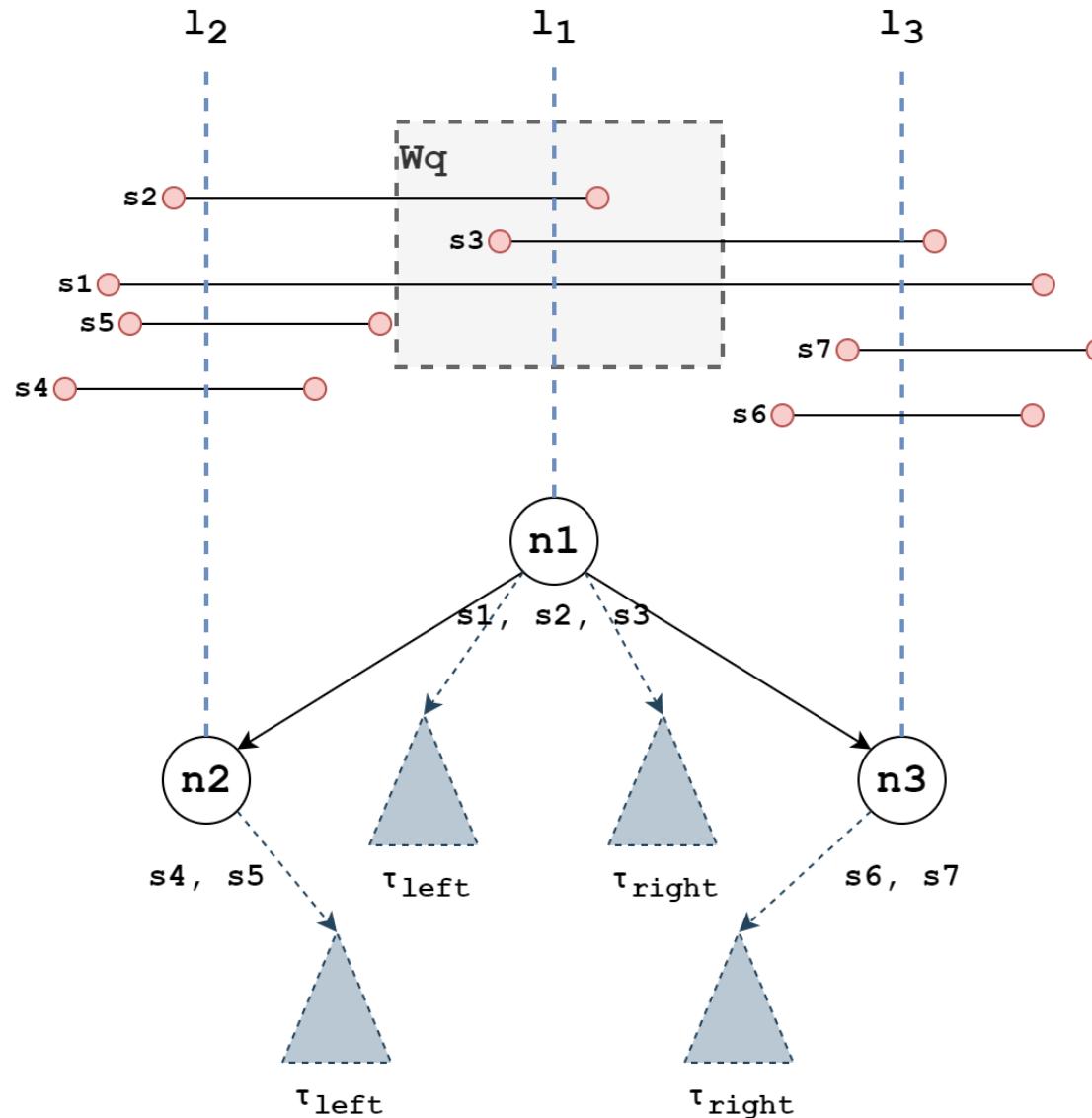
- Query window:  $W_q = [q_{x_1}, q_{x_2}] \times [q_{y_1}, q_{y_2}]$
- Similar to range tree search
- 3 cases:

if  $med(n_i) \in [q_{x_1}, q_{x_2}]$   
-> Go to both  $\tau_{left}$  and  $\tau_{right}$

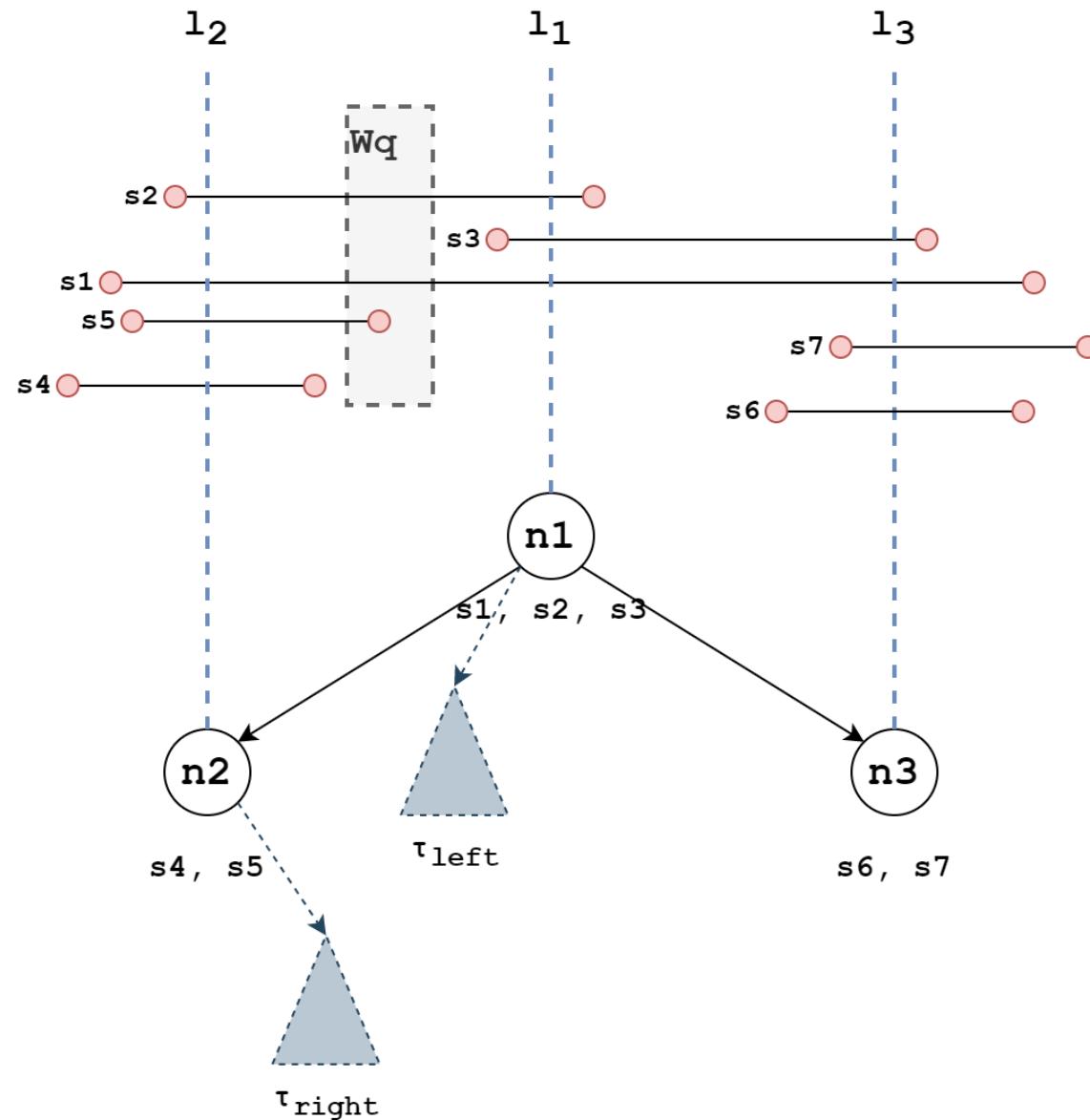
if  $med(n_i) > [q_{x_1}, q_{x_2}]$   
-> Search  $\tau_{right}$  for intervals right-ending in  $W_q$   
-> Skip  $\tau_{left}$

if  $med(n_i) \leq [q_{x_1}, q_{x_2}]$   
-> Search  $\tau_{left}$  for intervals left-ending in  $W_q$   
-> Skip  $\tau_{right}$

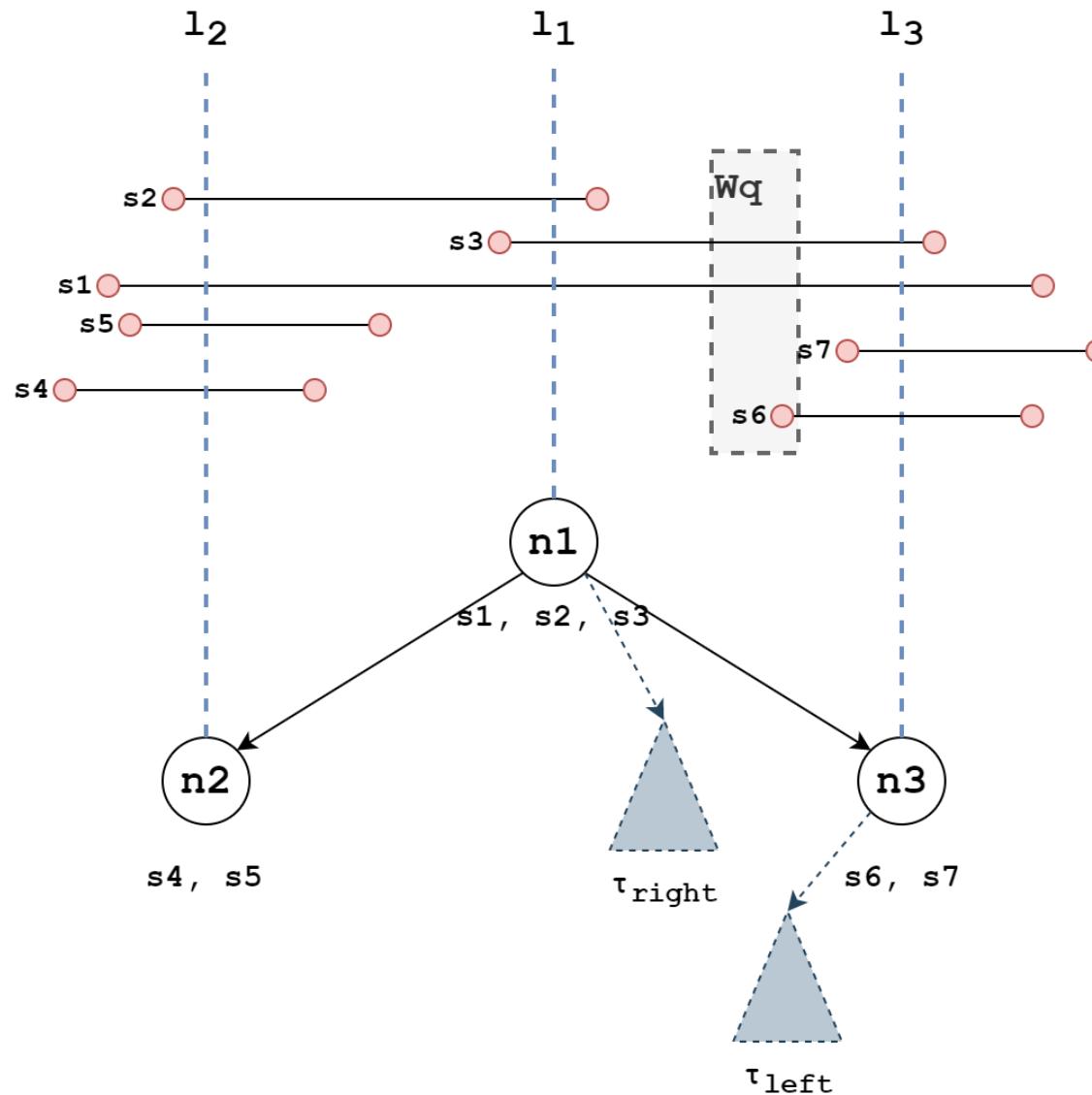
# Interval tree – search (2)



# Interval tree – search (3)

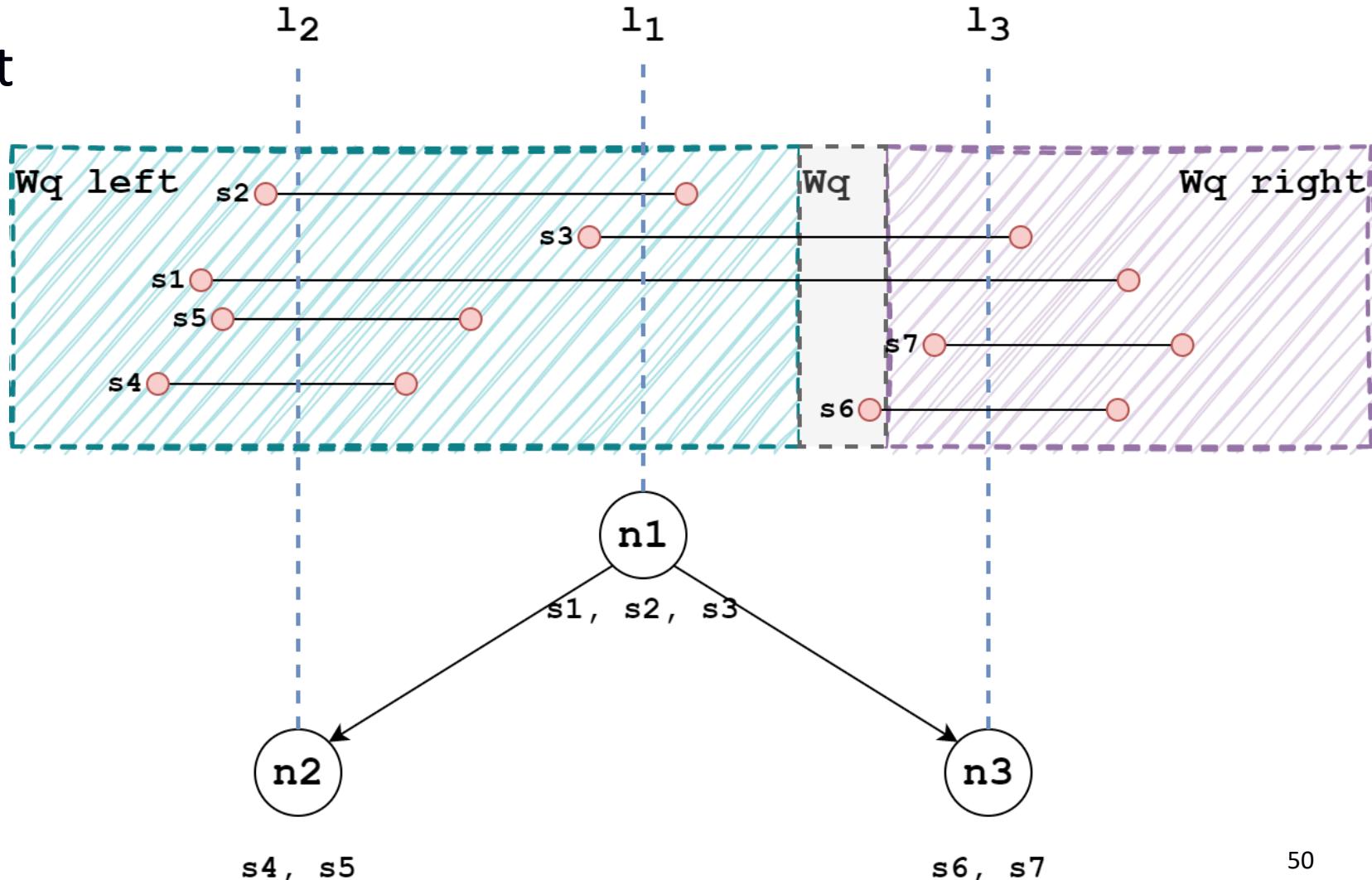


# Interval tree – search (4)



# Interval tree – search (5)

- Full intersect without endpoints?
- Search to infinity!

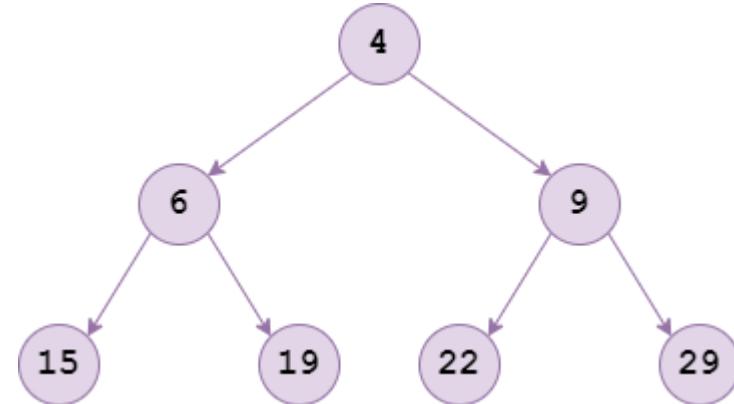


# Interval tree – search (6)

- 1<sup>st</sup> level tree:  $(\log n)$
- 2<sup>nd</sup> level 2D range trees:  $(2 \log^2 n)$
- Complete time complexity:  $O((2 \log^2 n)(\log n)) = O(\log^3 n)$
- Not that efficient?

# Priority tree (1)

- Use a heap
  - Sorted structure where the root is the greatest or least valued member
- Use an ascending heap
  - Sort the elements ascendingly
  - First element is the root
  - Left child of  $i^{\text{th}}$  node is at  $(2i + 1)^{\text{th}}$  index.
  - Right child of  $i^{\text{th}}$  node is at  $(2i + 2)^{\text{th}}$  index.
  - Parent of  $i^{\text{th}}$  node is at  $(i-1)/2$  index.



# Priority tree (2)

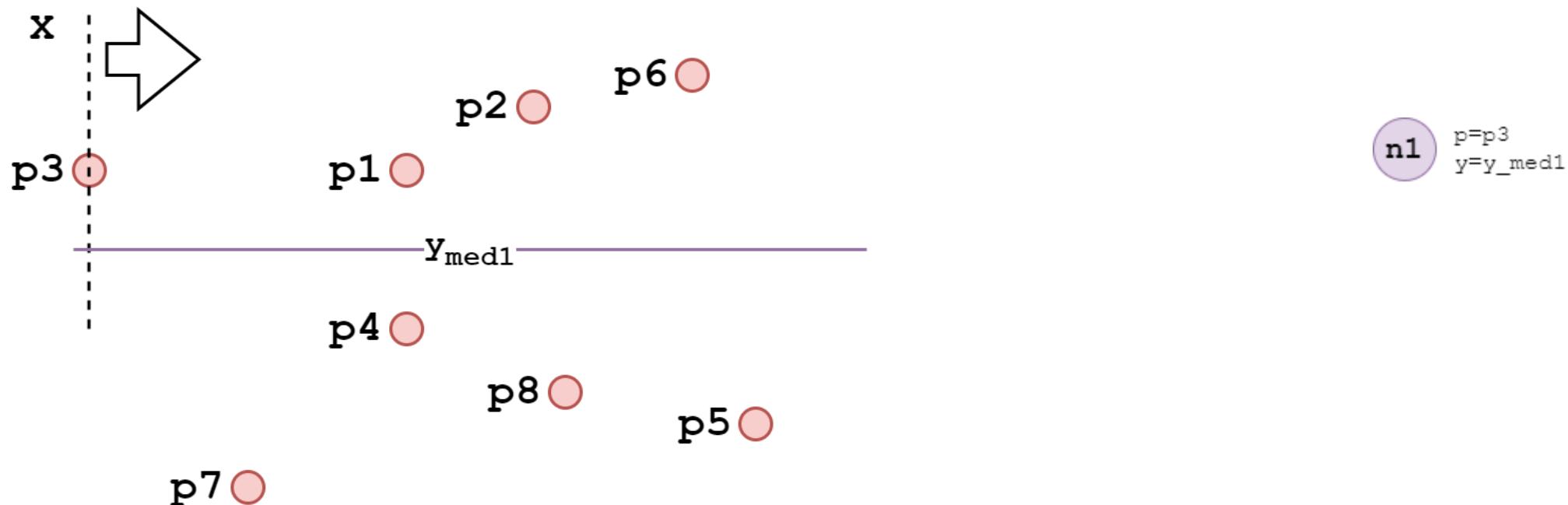
- Sort the points by  $p_x$
- Store the points in a heap
- Better searchability for  $(-\infty, q_{x_2}]$  and  $[q_{x_1}, +\infty)$ 
  - Depending on an ascending or descending heap
  - We hit “ $-\infty$ ” or “ $+\infty$ ” first – at the root

# Priority tree – creation (1)

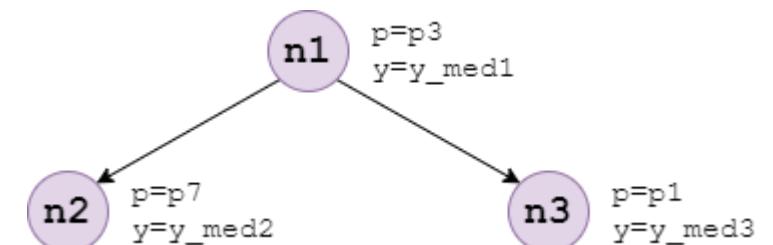
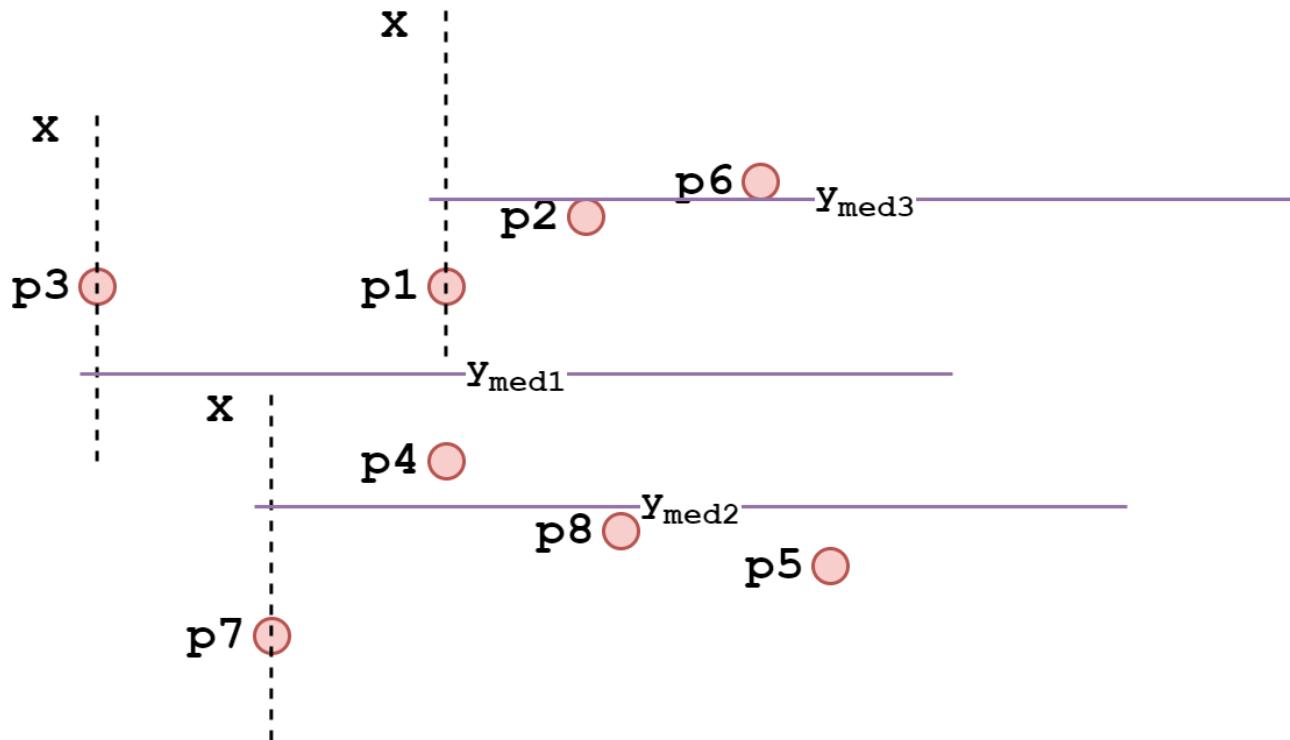
```
fun create_priority_tree(points):
    p_min = min_by_x(points)
    node = Node()
    node.point = p_min
    if points - p_min != []:
        y_med = median(points - p_min)
        points_left = [for p in points if p <= y_med]
        points_right = [for p in points if p > y_med]
        node.y = y_med
        if points_left != []:
            node.left = create_priority_tree(points_left)
        if points_right != []:
            node.right = create_priority_tree(points_right)
    node
```

# Priority tree – creation (2)

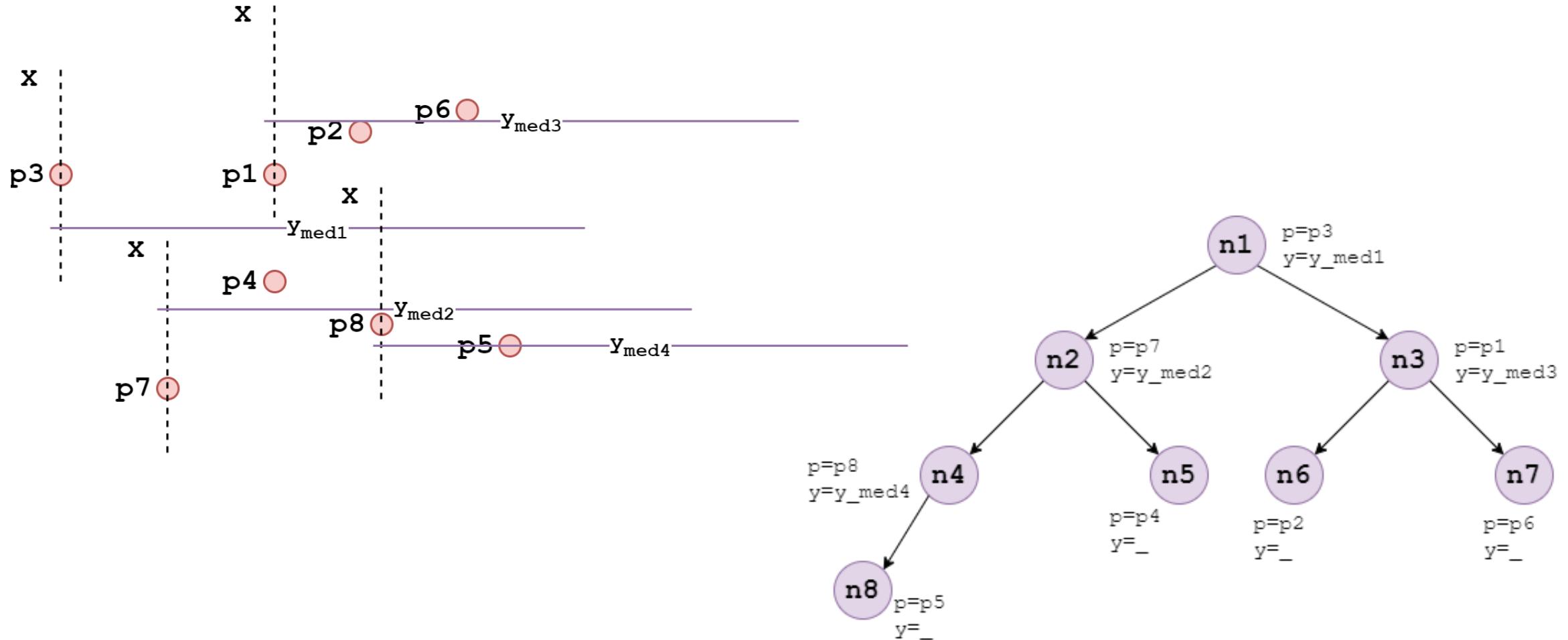
- We split the y-axis for every point along the points sorted by x



# Priority tree – creation (3)



# Priority tree – creation (4)



# Priority tree – search (1)

- The main problem in the priority tree:

node.point.y != node.y\_med

- They can differ greatly, meaning:

node.point.y in [q\_y1, q\_y2]

node.y\_med not in [q\_y1, q\_y2]

- Path finding uses binary tree part of the priority tree: node.y\_med
- Heap uses points stored in the tree nodes: node.point

# Priority tree – search (2)

- The main problem in the priority tree:

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node.point.y in [q\_y1, q\_y2]

node.y\_med not in [q\_y1, q\_y2]

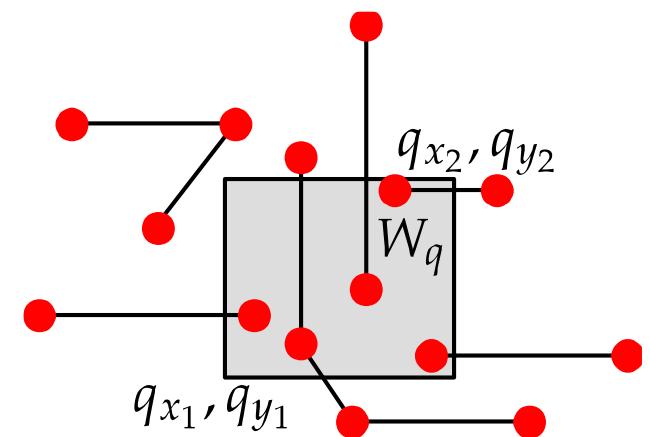
- Path finding uses binary tree part of the priority tree: node.y\_med
- Heap uses points stored in the tree nodes: node.point

# Priority tree – search (3)

- Search for the splitting node by `node.y_med` in  $[q_{y_1}, q_{y_2}]$ 
  - 2 vertical tree traverses
- Continually check that `node.point.x ≤ q_{x_2}`
- We are checking for points in  $(-\infty, q_{x_2}] × [q_{y_1}, q_{y_2}]$ 
  - Ascending heap

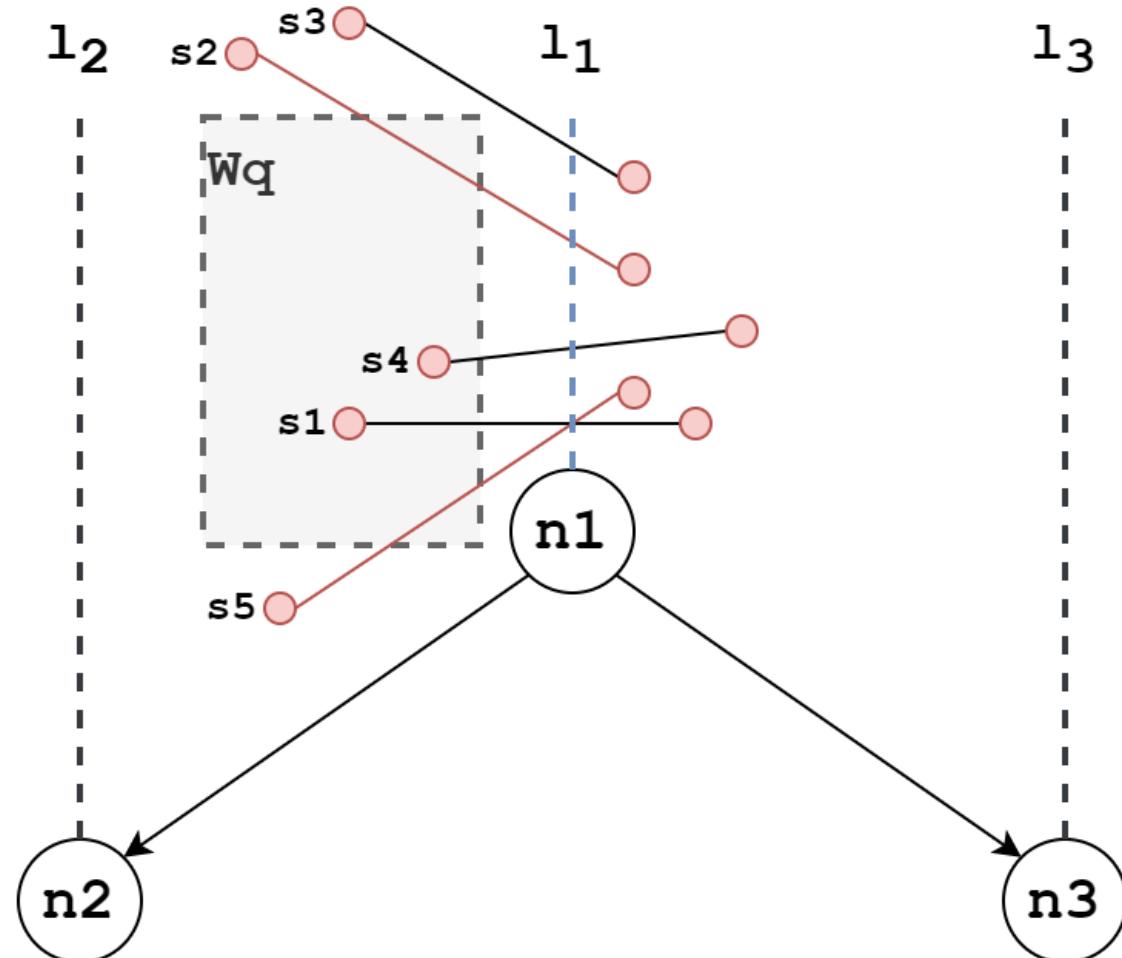
# Line segment search (1)

- Line segment set:  $S = \left\{ s_i = \overline{(x_{i_1}, y_{i_1})(x_{i_2}, y_{i_2})} : (x_{i_k}, y_{i_k}) \in \mathbb{R}^2 \right\}$
- Search area:  $W_q = [q_{x_1}, q_{x_2}] \times [q_{y_1}, q_{y_2}]$



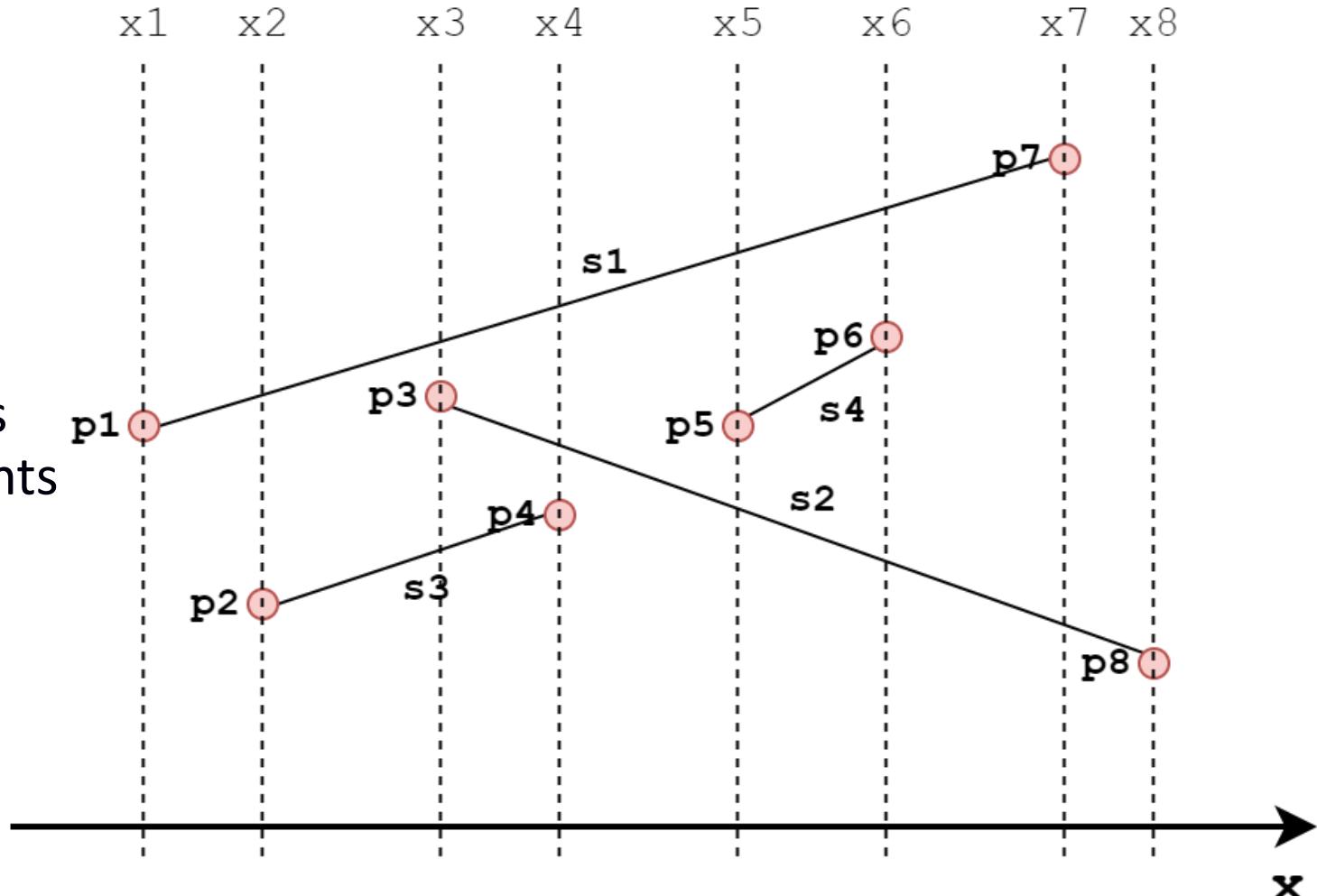
# Line segment search (2)

- Interval tree?
- What's wrong with **s2** and **s5**?
- False negative



# Line segment search (3)

- Locus approach
- Mark elementary intervals
  - Each interval continually has the same number of segments

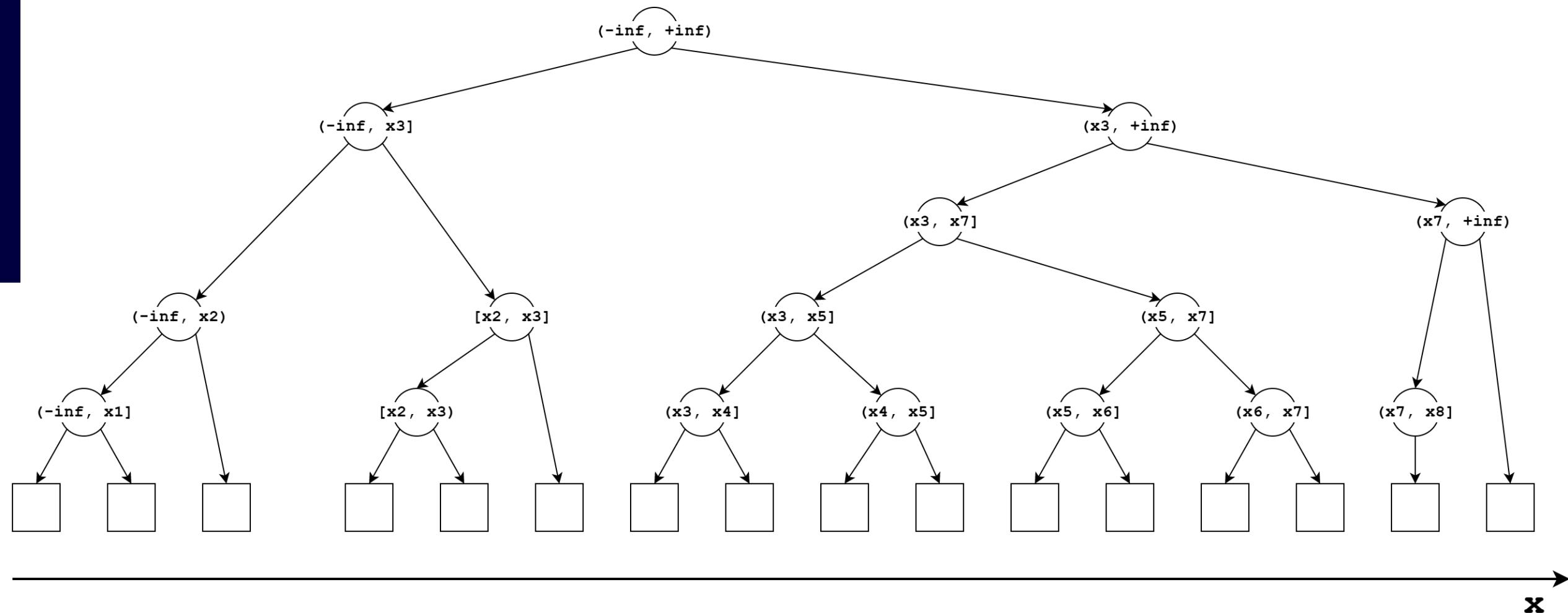


$(-\infty, x_1), [x_1, x_1], (x_1, x_2), [x_2, x_2], \dots, (x_7, x_8), [x_8, x_8], (x_8, \infty)$

# Line segment search (4)

- Each interval becomes a tree node in a balanced index tree
  - We could use the B+ tree
- Leaves are sets of line segments passing through the interval
  - A line segment can be in multiple leaves (intervals)
- Parent nodes aggregate child node intervals:
$$I(n) = I(n_L) \cup I(n_R)$$
- Search complexity:  $O(k + \log^2 n)$

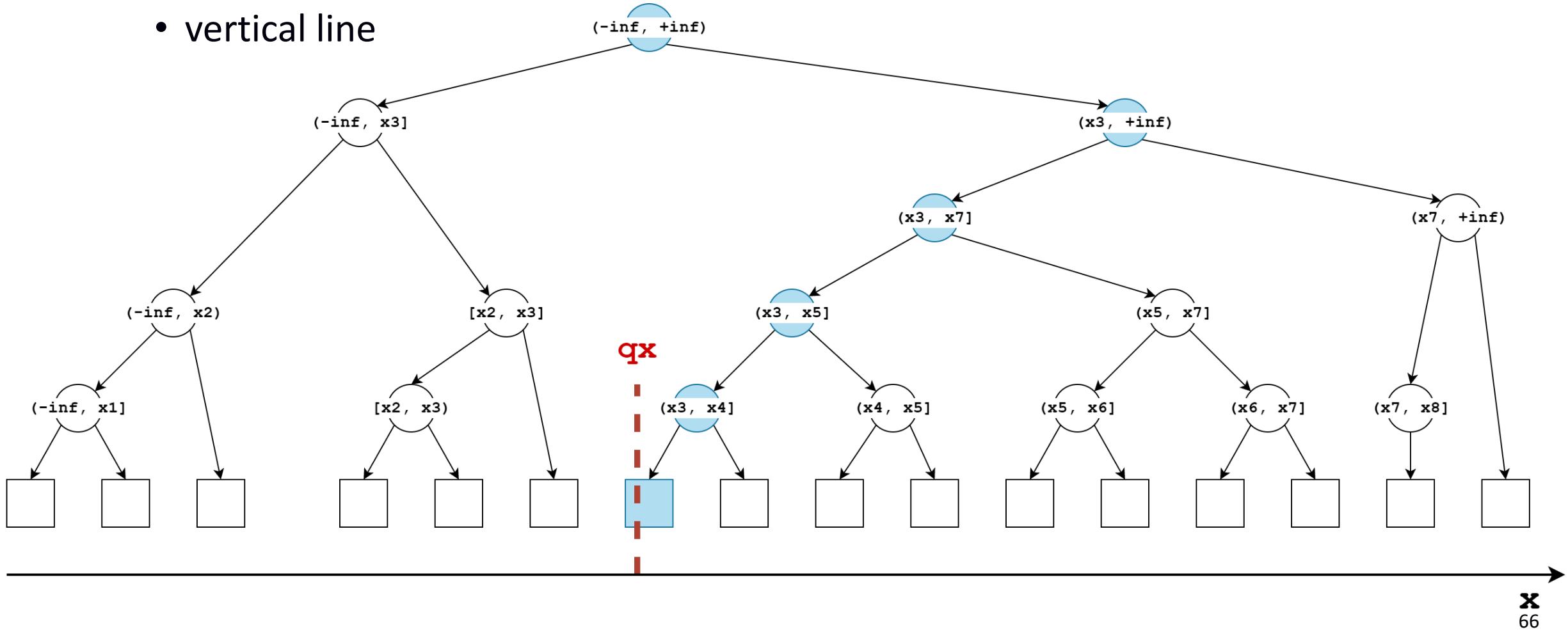
# Line segment search (5) – segment tree



# Line segment search (6)

- We can search for segments going through  $q_x \times [q_{y1}, q_{y2}]$

- vertical line



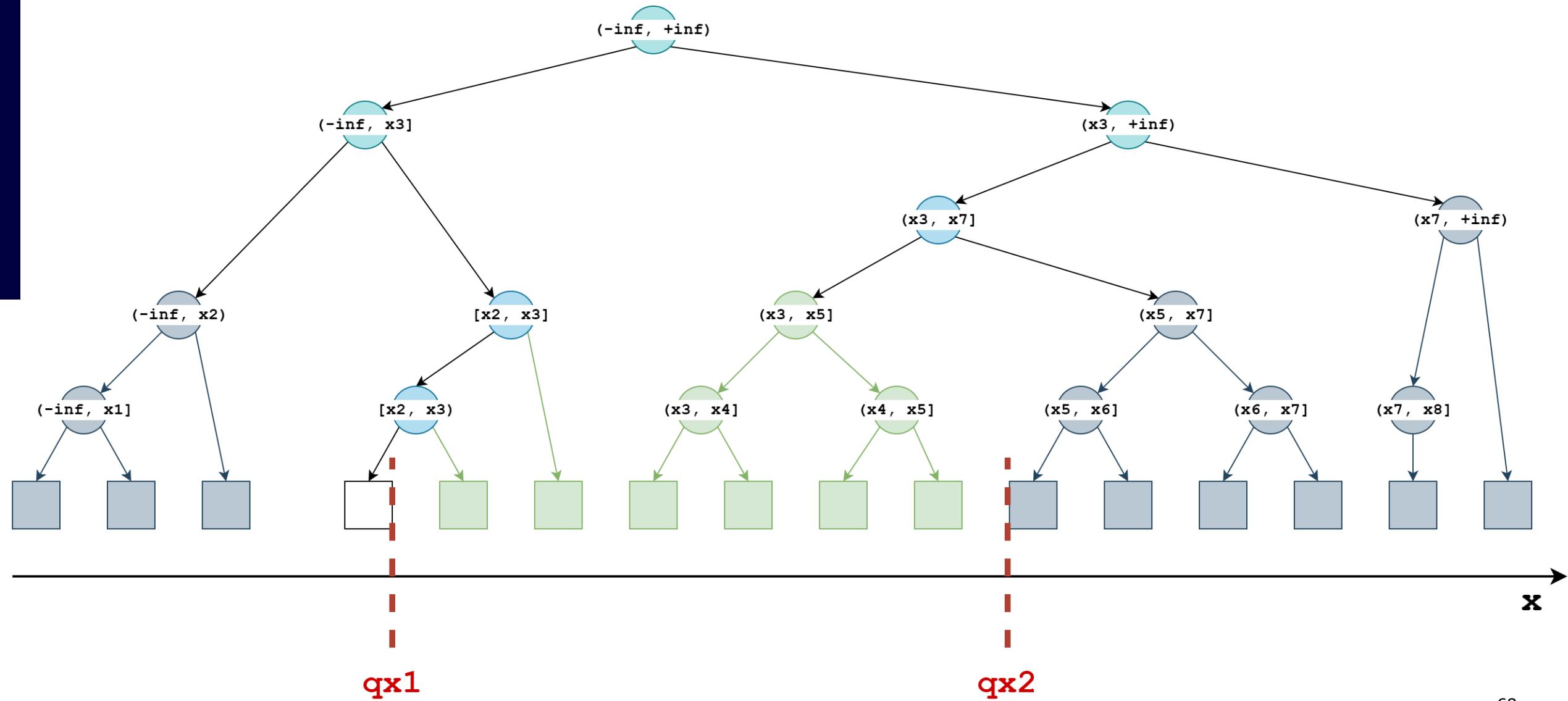
# Line segment search (7)

- We can search for segments going through

$$W_q = [q_{x_1}, q_{x_2}] \times [q_{y_1}, q_{y_2}]$$

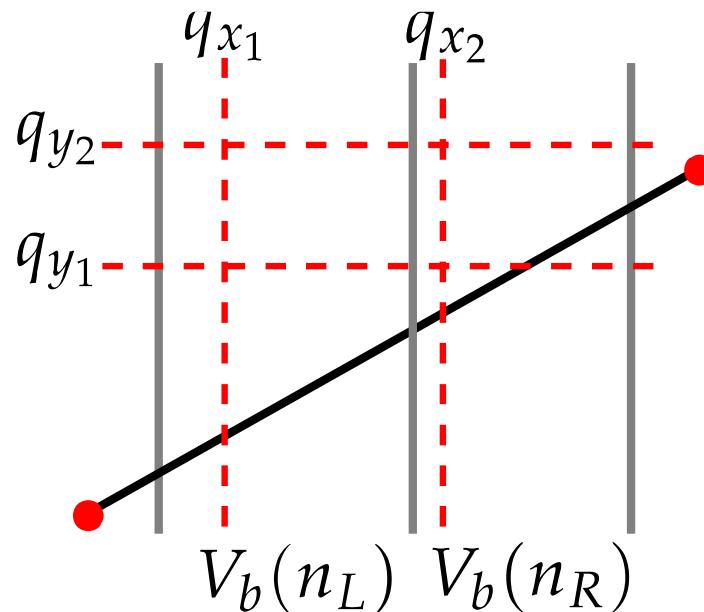
- Again we are searching for  $n_{split}$
- We proun or skip or traverse down the tree
- The result is a set of vertical blocks (unbounded by y)

# Line segment search (8)



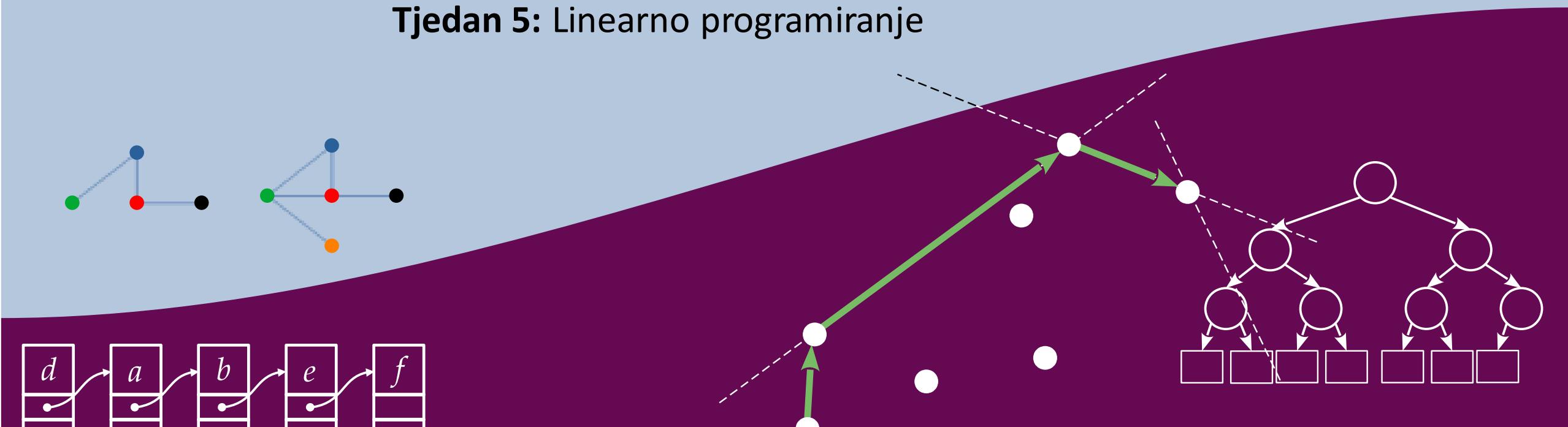
# Line segment search (9)

- What about  $[q_{y_1}, q_{y_2}]$ ?
- Another embedded tree?
  - Range search?
- Crossed lines?
  - Problem for verticality
- Out of bounds segment?
  - Line segment does enter vertical box, but “after” the search box bounds
  - Possible *false positive*



# Napredni algoritmi i strukture podataka

Tjedan 5: Linearno programiranje

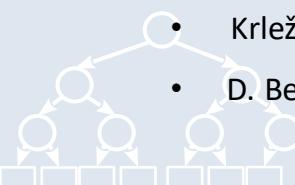


# Što je linearni program (LP)?

- Dosta **generalna** klasa problema koja se može rješavati efikasno
- Brojne primjene u industriji
  - Lanci nabave
  - Raspoređivanje
  - Optimizacije u električnim mrežama
- Spada u kategoriju **konveksnih optimizacijskih problema**
  - LP je prva podkategorija koja je efikasno riješena (cca 1940tih)
  - Ostale kategorije su slijedile kasnije, otvoreno područje istraživanja

## Literatura:

- Krleža, Brčić: „Advanced Algorithms and Data Structures”, Lecture book. (4. poglavlje)
- D. Bertsimas, and J. Tsitsiklis.: Introduction to linear optimization. Athena Scientific, 1997. (poglavlja 3.1-3.3, 3.5)



# Što je linearni program (LP)?

- Rješavanje linearnih programa je dignuto na razinu industrijske pouzdanosti
  - Pouzdani i brzi alati
    - Gurobi – trenutno najbrži rješavač
    - Python – `scipy.optimize.linprog`
    - Čak integrirani u Excel
  - Bitni i za dizajn i analizu algoritama, npr.
    - Algoritmi nad grafovima
    - Približni algoritmi



# Linearni program (LP)?! Generalno...

minimizirati

$$\mathbf{c}^T \mathbf{x}$$

uz uvjet

$$\mathbf{A}\mathbf{x} \leq \mathbf{b}$$

$$\mathbf{D}\mathbf{x} = \mathbf{e}$$

pri čemu je:  $\mathbf{x} \in \mathbb{R}^n$ ,  $\mathbf{c} \in \mathbb{R}^n$ ,  $\mathbf{b} \in \mathbb{R}^m$ ,  $\mathbf{e} \in \mathbb{R}^k$ ,

matrica  $\mathbf{A} \in \mathbb{R}^{m \times n}$ , matrica  $\mathbf{D} \in \mathbb{R}^{k \times n}$

- Ciljna funkcija  $f$  (*objective function, cost function*)
- Varijable odluke  $\mathbf{x}$  (*control, structural, decision variables*)
- Ograničenja sa parametrima  $\mathbf{A}, \mathbf{b}, \mathbf{D}, \mathbf{e}$  (*constraints*)



# Primjer I – humanitarni transportni problem

- Pfizer proizvodi cjepiva za COVID-19 te ima proizvodne pogone u tri mesta T1...3 s raspoloživim kapacitetima zadanim u tablici. Naručitelji su iz četiri mesta O1...4 sa potrebama zadanim u zadnjem retku tablice. Jedinični transportni troškovi za sve kombinacije proizvodnih pogona i naručitelja su navedeni u tablici.
- Kako na najefikasniji način zadovoljiti potrebe naručitelja?



	Transportni troškovi				
	O1	O2	O3	O4	Kap.
T1	10	9	14	8	432
T2	7	11	9	11	138
T3	8	12	12	9	35
Nar.	500	200	115	100	5/58

# Primjer I – humanitarni transportni problem

$x_{ij}$  - količina isporučenu iz  $i$ -te tvornice  $j$ -tom naručitelju

$c_{ij}$  - trošak transporta po jedinici između  $i$  i  $j$

$$\min \left( \sum_{i,j} x_{ij} c_{ij} \right)$$

uz uvjete  $\sum_j x_{ij} \leq s_i \quad ; \quad i = 1, 2, 3$

$$\sum_i x_{ij} = d_j \quad ; \quad j = 1, 2, 3, 4$$
$$x_{ij} \geq 0$$



# Primjer II – optimalno uparivanje u online datingu

Na dating siteu u hetero rubrici postoji  $M$  muškaraca i  $F$  žena. Na temelju popunjenih upitnika raznim modelima su izračunate kompatibilnosti za sve potencijalne parove. Site želi upariti sve u parove da bi ukupna suma normaliziranih kompatibilnosti bila što veća (utilitarizam).

## Kombinatorni problem

Naivno rješenje: ispitati sve kombinacije F-M. Treba ispitati  $F!$  kombinacija (ako je  $F=M$ )

# Primjer II – optimalno uparivanje u online datingu

$x_{ij}$  - 1 ako je  $i$  uparen sa  $j$ , 0 inače

$k_{ij}$  – kompatibilnost za uparivanje  $(i,j)$

$$\max \left( \sum_{i,j} x_{ij} k_{ij} \right)$$

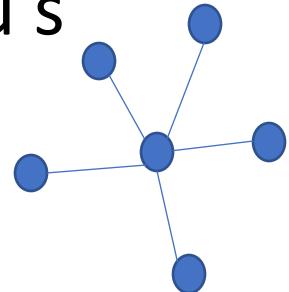
uz uvjete  $\sum_j x_{ij} = 1, \forall i = 1, \dots, F$

$$\sum_i x_{ij} = 1, \forall j = 1, \dots, M$$
$$x_{ij} \geq 0$$



# Primjer III – bežične raspodijeljene mreže

Centralna postaja treba ostvariti pouzdanu bežičnu komunikaciju s  $N$  raspodijeljenih senzora za praćenje klimatskih promjena.



Senzori se napajaju Sunčevom energijom i važno je smanjiti njihovu potrošnju. Signal  $i$ -tog senzora do centralne postaje stiže **prigušen**. Ako  $i$ -ta senzor emitira snagom  $p_i$ , centralna postaja prima signal snage  $\lambda_i \cdot p_i$  ( $\lambda < 1$ ). Tijekom komunikacije s  $i$ -tim senzorom, signali svih drugih senzora koji dolaze u centralnu postaju predstavljaju smetnju i komunikacija je moguća samo ako je omjer signal/šum najmanje  $\rho_i$ .

Kolike trebaju biti emitivne snage  $p_i$  senzora kako bi se ostvarila pouzdana komunikacija uz najmanju moguću potrošnju energije?



# Primjer III – bežične raspodijeljene mreže

Ukupna potrošnja je proporcionalna ukupnoj snazi pa ćemo to minimizirati:

$$\min \left( \sum_{i=1}^N p_i \right)$$

uz uvjete

$$\frac{\lambda_i p_i}{\sum_{j \neq i} \lambda_j p_j} \geq \rho_i \quad ; \quad i, j = 1, \dots, N$$

$$p_k \geq 0 \quad ; \quad k = 1, \dots, N \quad .$$

Budući da uvjetne (ne)jednadžbe moraju biti linearne po  $p_k$ , prevodimo ih u oblik

$$\lambda_i p_i - \rho_i \sum_{j \neq i} \lambda_j p_j \geq 0$$



# LP formulacije

- Općenita formulacija je „neuredna”
- Dvije formulacije kojima težimo radi lakšeg rješavanja i pisanja algoritama
  - **Kanonska forma LP** – idealna za geometrijsku perspektivu
  - **Standardna forma LP** – idealna za algebarsku perspektivu
- Svi ostali LP se mogu prevesti u obje forme\*

\*pročitajte o transformacijama u skripti



# Kanonska forma LP

minimizirati  
uz uvjet

$$\begin{aligned} & \mathbf{c}^T \mathbf{x} \\ & \mathbf{Ax} \leq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

pri čemu je:  $\mathbf{c} \in \mathbb{R}^n$ ,  $\mathbf{b} \in \mathbb{R}^m$ , matrica  $\mathbf{A} \in \mathbb{R}^{m \times n}$ .



# Standardna forma LP

minimizirati  
uz uvjet

$$\begin{aligned} & \mathbf{c}^T \mathbf{x} \\ & \mathbf{Ax} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

pri čemu je:  $\mathbf{c} \in \mathbb{R}^n$ ,  $\mathbf{b} \in \mathbb{R}^m$  i  $\mathbf{b} \geq \mathbf{0}$ , matrica  $\mathbf{A} \in \mathbb{R}^{m \times n}$

Nije dio definicije, ali prepostavljat ćemo u sklopu predavanja da je  $\text{rang}(\mathbf{A}) = m$  i  $m < n$ . Ako je rang manji, linearno-zavisna ograničenja se mogu ukloniti.



# Grafička metoda - primjer

$$\max \quad x_1 + 5 \cdot x_2$$

uz uvjete

$$\begin{bmatrix} 5 & 6 \\ 3 & 2 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 30 \\ 12 \end{bmatrix}$$

$$x \geq 0$$

[GeoGebra](#) – interaktivna geometrija



# Grafička metoda - primjer

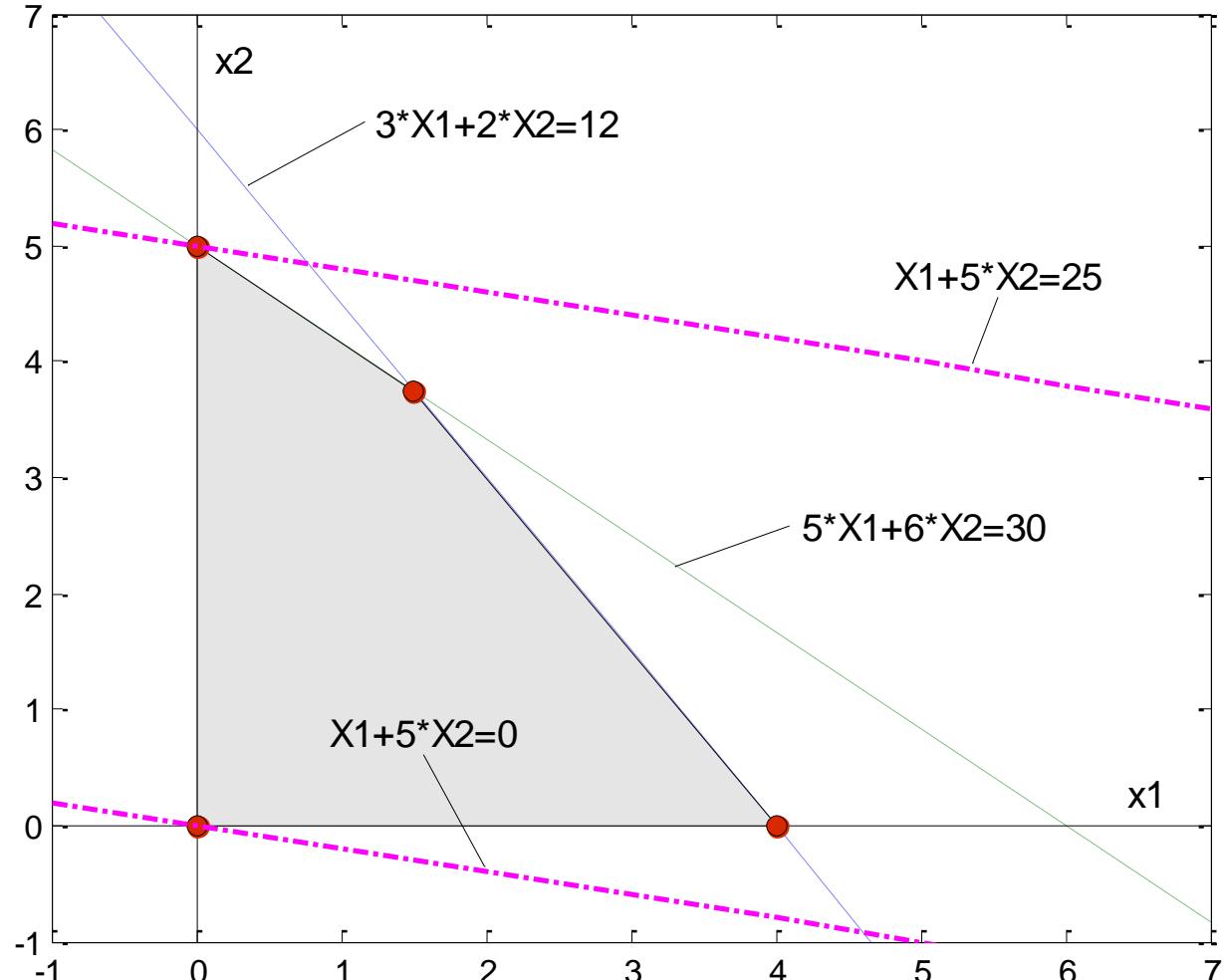
$$\begin{array}{ll}\text{max} & x_1 + 5 \cdot x_2 \\ \text{uz uvjete} & \begin{bmatrix} 5 & 6 \\ 3 & 2 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 30 \\ 12 \end{bmatrix} \\ & x \geq 0\end{array}$$

Rješenje je sjecište pravca  $x_1 + 5 \cdot x_2 = f$  s prostorom svih mogućih rješenja (sivi poligon na slici) za koje je  $f$  najveća.

## Rješenje

$$x = [0, 5]^T$$

$$f_{\max} = 25$$

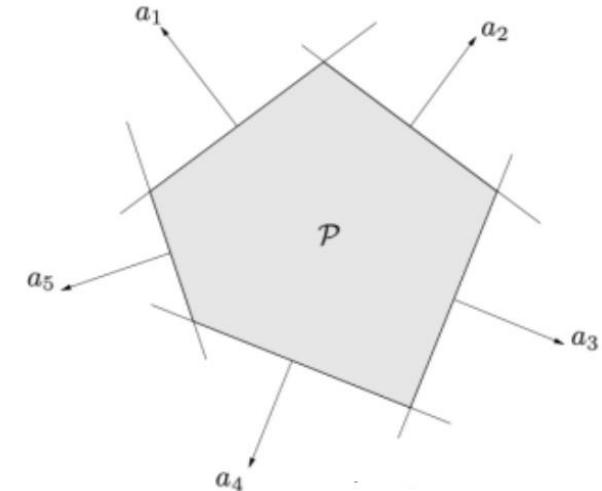
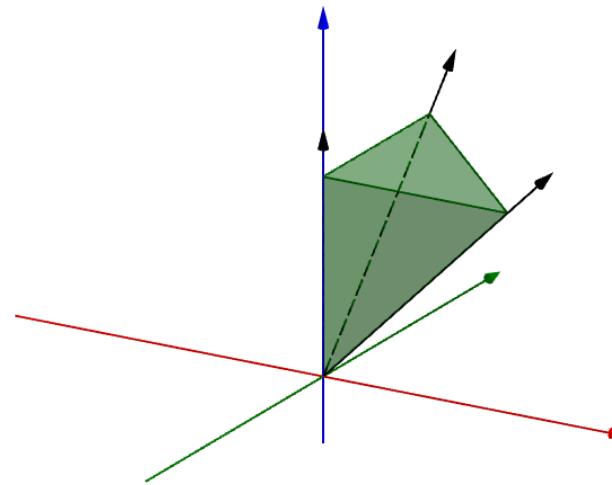
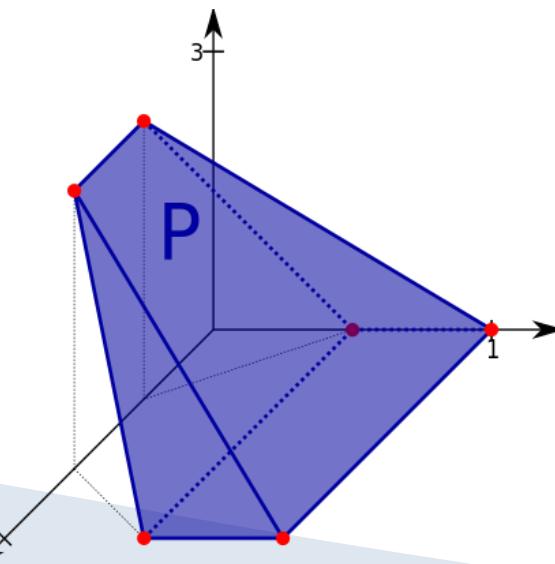


# Geometrijska analiza

**Definicija.** *Konveksni politop* u n-dimenzionalnom prostoru jest skup vektora (točaka):

$$\{x \in \mathbb{R}^n \mid Ax \leq b\}$$

- Ograničenja linearnih programa!



# Geometrijska analiza

Tijela opisana ograničenjima u LP su konveksni politopi P

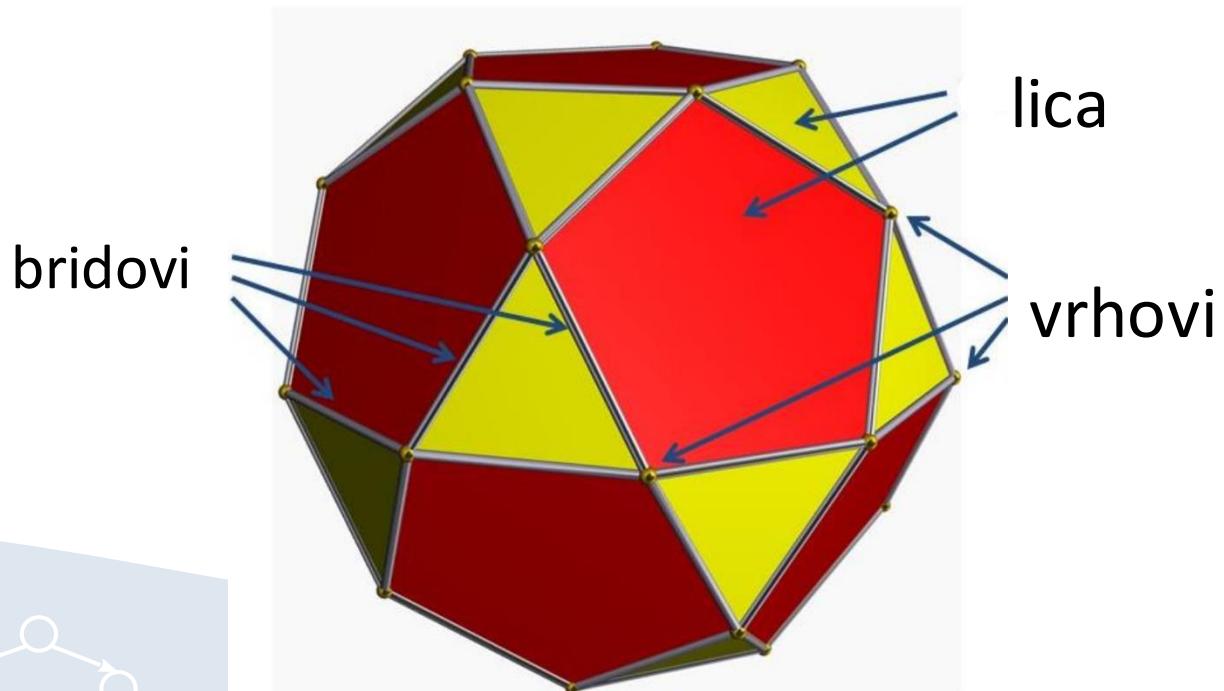
$$P = \{x \in \mathbb{R}^n \mid Ax \leq b\}$$



# Geometrijska analiza

Tijela opisana ograničenjima u LP su konveksni politopi P

$$P = \{x \in \mathbb{R}^n \mid Ax \leq b\}$$



**Definicija.** Aktivno ograničenje u nekoj točki  $x$  je svako ograničenje koje je ispunjeno sa jednakosti u toj točki  $x$ .

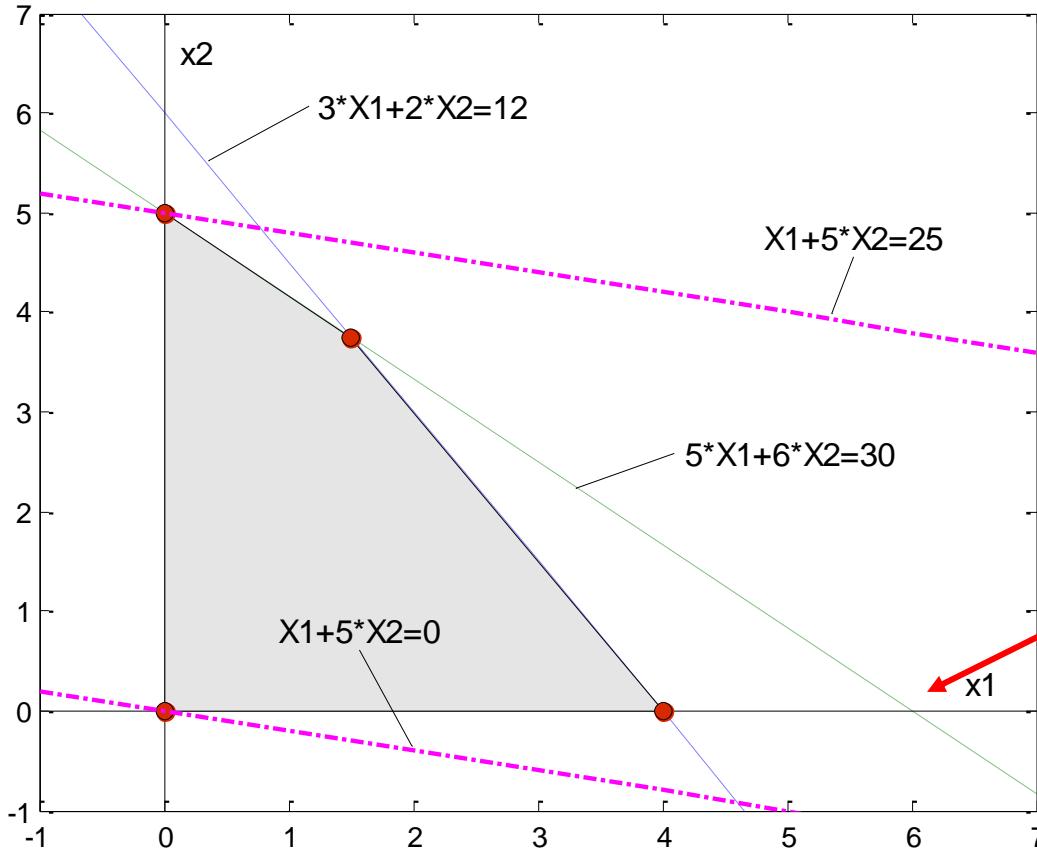
**Definicija.** U  $n$ -dimenzionalnom prostoru, vrh politopa je definiran kao presjecište barem  $n$  aktivnih linearne nezavisnih ograničenja pri čemu su ostala ograničenja zadovoljena.



# Geometrijska analiza

Tijela opisana ograničenjima u LP su konveksni politopi P

$$P = \{x \in \mathbb{R}^n \mid Ax \leq b\}$$



**Definicija.** U  $n$ -dimenzionalnom prostoru, **vrh politopa** je definiran kao presjecište barem  $n$  aktivnih linearne nezavisnih ograničenja pri čemu su ostala ograničenja zadovoljena.

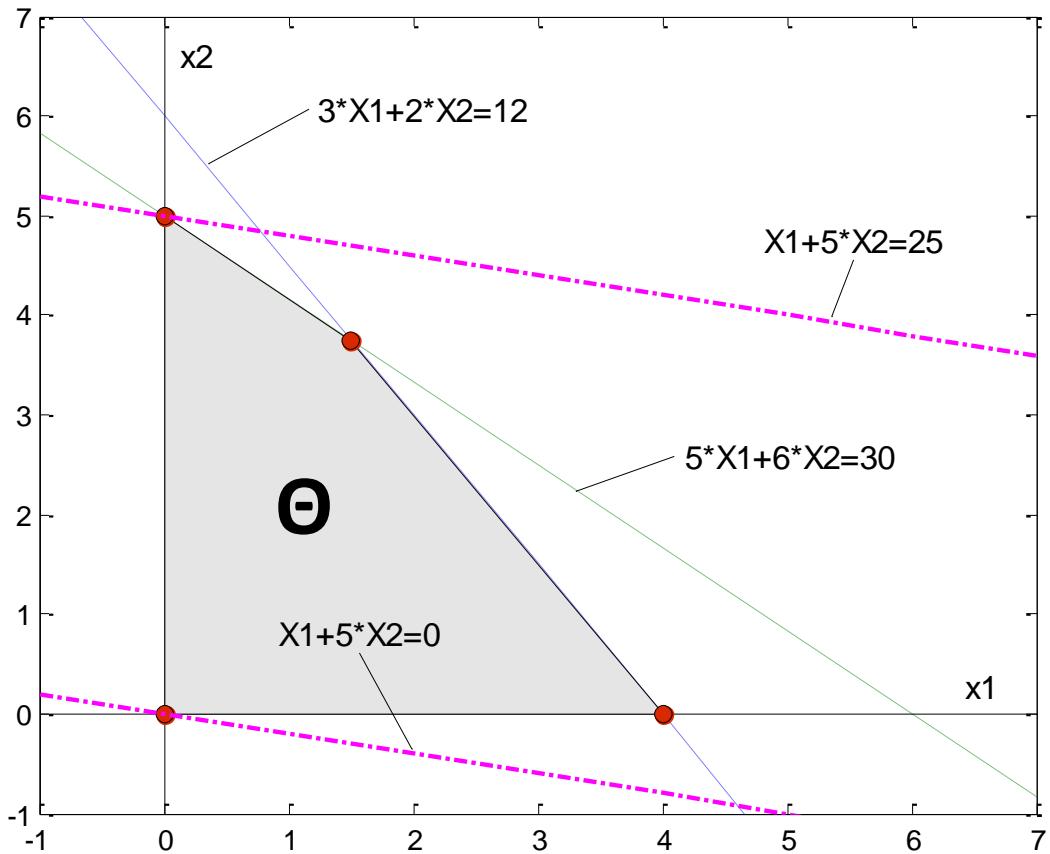
**Oprez!** Nije svako presjecište  $n$  aktivnih ograničenja **vrh**! Neaktivna ograničenja moraju biti ispoštovana da bismo bili u vrhu.

Susjedi?

19/58

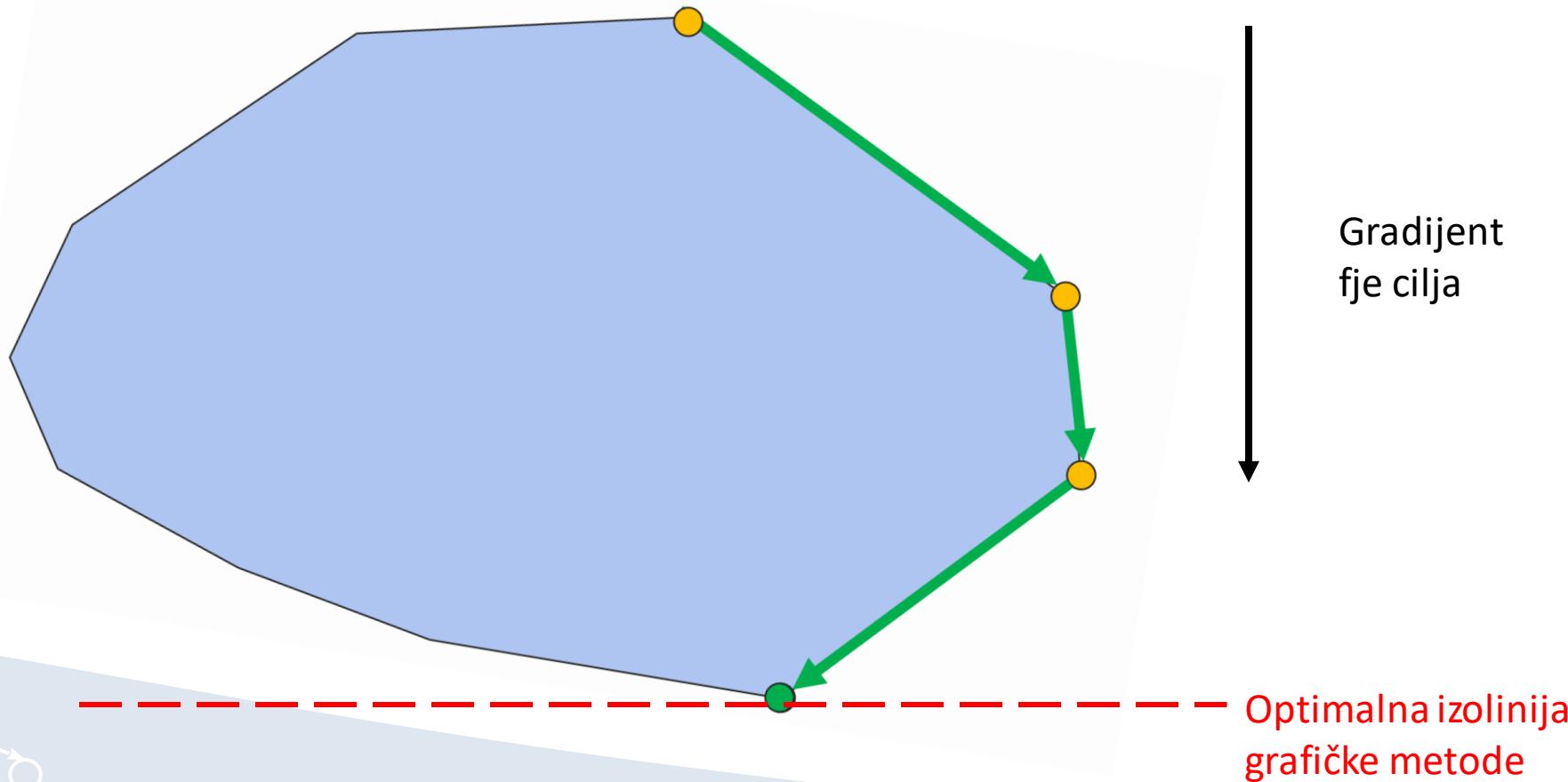
# Geometrijska analiza

- Definicija. Skup  $\Theta$  je **konveksni skup** ako sadrži sve točke ravne spojnice između bilo kog para točaka iz skupa  $\Theta$ .  
 $\forall x, y \in \Theta, \forall \alpha \in (0, 1): \alpha x + (1 - \alpha)y \in \Theta$



- Definicija. **Ekstrem konveksnog skupa**  $\Theta$  je svaka točka  $x \in \Theta$  koja nije na ravnoj spojnici ikojih drugih dviju točaka skupa  $\Theta$ .  
 $(\exists x_1, x_2 \in \Theta \setminus \{x\})(\nexists \alpha \in (0, 1))$   
$$x = \alpha x_1 + (1 - \alpha)x_2$$
- Ekstremi u politopu su **vrhovi** – geometrijski koncept

# Simplex - ideja



# Simplex – ulazni problem

Za simpleks koristimo **LP u standardnoj formi**

minimizirati

$$\mathbf{c}^T \mathbf{x}$$

uz uvjet

$$\mathbf{A}\mathbf{x} = \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

pri čemu je:  $\mathbf{c} \in \mathbb{R}^n$ ,  $\mathbf{b} \in \mathbb{R}^m$  i  $\mathbf{b} \geq 0$ , matrica  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rang}(\mathbf{A}) = m$  i  $m < n$

Imamo **m** ograničenja jednakosti, **n** ograničenja nejednakosti ( $\mathbf{x} \geq \mathbf{0}$ )

Nalazimo se u **n**-dimenzionalnom prostoru ( $n \geq m$ )

**Vrhovi** određeni sa **n** aktivnih ograničenja (oprez!)



# Simplex – jezgra metode

- Nalazimo se u **n**-dimenzionalnom prostoru ( $n \geq m$ )
  - Vrhovi određeni sa **n aktivnih ograničenja**
  - **m aktivnih ograničenja je već fiksirano**
  - „**Proizvoljnih**“ ( $n-m$ ) biramo među nejednakostima
    - **One fiksiraju vrijednosti ( $n-m$ ) varijabli na 0**
    - Te varijable ćemo nazivati **nebazičnima**
    - Kad ih uvrstimo u  $m$  ograničenja, dobijemo sustav  $m$  jednadžbi sa  $m$  nepoznanica! (znamo rješiti iz linearne algebре) – varijable koje rješavamo nazivamo **bazične**
  - Particionira se skup svih varijabli na **bazične i nebazične**



# Simplex – definicije

**Definicija.** **Bazično rješenje** sustava  $\mathbf{Ax} = \mathbf{b}$ ,  $\mathbf{x} \geq \mathbf{0}$  je vektor  $\mathbf{x} = [\mathbf{x}_B^T \ \mathbf{0}]$ , gdje je  $\mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b}$ , a  $\mathbf{B} \in \mathbb{R}^{m \times m}$  odabrana baza (stupci u matrici  $\mathbf{A}$ ) u sustavu od  $m$  jednadžbi s  $n$  nepoznanica, pri čemu je  $m < n$  i  $\det(\mathbf{B}) \neq 0$ .

## Teorem (osnovni teorem linearog programiranja):

Promatrajmo linearni problem u standardnoj formi. Vrijedi sljedeće:

1. Ako postoji bilo kakvo rješenje, postoji i **izvedivo bazično rješenje**.
2. Ako postoji optimalno rješenje, postoji i bazično optimalno rješenje.



# Tabličenje

Hajdemo staviti sve parametre u tablicu:

- $m+1$  redaka
- $n+1$  stupaca

$c^T$		0
	$A$	$b$



# Tabličenje

- Particioniranje A po stupcima na bazične B i nebazične N stupac-vektore

$c_B^T$	$cN^T$	0
$B$	$N$	$b$



# Simplex tableau

Iterativno implicitno izračunavanje inverza  $B^{-1}$

- Efikasnije izvođenje jer se susjedni vrhovi razlikuju samo u **jednom aktivnom ograničenju** (koje postavlja neku drugu varijablu na 0)

		faktori redukcije po bridovima	-fja cilja
		$\mathbf{c}_N^T - \mathbf{c}_B^T B^{-1} \mathbf{N}$	$-\mathbf{c}_B^T B^{-1} \mathbf{b}$
		$B^{-1} \mathbf{N}$	$B^{-1} \mathbf{b}$
$0^T$			
$I$			
		bridovi do susjednih baz.rješ	Vrijednosti baz.var.



# Simplex – pseudokod

1. **Početak** iz izvedivog bazičnog rješenja u simpleks tablici
2. Optimalno? Ako su svi faktori redukcije nenegativni, **STOP**
3. Tranzicija u boljeg susjeda:
  - a) Odabir ulazne nebazične varijable koja odgovara stupcu  $q$
  - b) Odabir izlazne bazične varijable koja odgovara retku  $p$ . Ako ne postoji – problem je neograničen, **STOP**
  - c) Gauss-Jordan eliminacija za pivot  $(p,q)$
4. Povratak na korak 2



# Simplex – pseudokod

- Odabir ulazne nebazične varijable koja odgovara stupcu  $q$ 
  - Odabere se neka koja ima **NEGATIVNI** faktor redukcije
- Odabir izlazne bazične varijable  $x_{[p]}$  koja odgovara retku  $p$ 
  - $p = \operatorname{argmin}_{i \in \{1, \dots, m\}} \{x_{[i]} / B^{-1}A_{iq} \mid B^{-1}A_{iq} > 0\}$

\*[p] označava odabir varijable preko reference retkom



# Simplex – primjer

$$\begin{aligned} \max \quad & 7x_1 + 6x_2 \\ \text{uz} \quad & 2x_1 + x_2 \leq 3 \\ & x_1 + 4x_2 \leq 4 \\ & x_1, x_2 \geq 0 \end{aligned}$$

Uvođenjem dviju *slack* varijabli  $x_3$  i  $x_4$  prevodimo LP u standardnu formu

$$\begin{aligned} \min \quad & -7x_1 - 6x_2 + 0x_3 + 0x_4 \\ \text{uz} \quad & 2x_1 + x_2 + x_3 = 3 \\ & x_1 + 4x_2 + + x_4 = 4 \\ & x_1, x_2, x_3, x_4 \geq 0 \end{aligned}$$



# Simplex – primjer

Tablični zapis je

	$a_1$	$a_2$	$a_3$	$a_4$	b = RHS	
$c^T$	-7	-6	0	0	0	$= r^T$
	2	1	1	0	3	
	1	4	0	1	4	

- Tablica već valjana, vrijedi  $r_i = c_i$
- najlakše je odabratи почетну базу  $B_0 = [a_3 \ a_4] = I_2$ 
  - bazično rješenje  $x_{(0)} = [0 \ 0 \ 3 \ 4]^T, f(x_{(0)})=0$
- Nulti redak sadrži faktore redukcije
  - ima negativnih pa nije optimum



# Simplex – primjer

Tablični zapis je

	$a_1$	$a_2$	$a_3$	$a_4$	RHS	
$c^T$	-7	-6	0	0	0	$= r^T$
	2	1	1	0	3	
	1	4	0	1	4	

Nulti redak ima negativne faktore redukcije pa nije optimum!

Odabiremo  $q=1$  (prvi stupčani vektor,  $a_1$ , ulazi u bazu)

Odabiremo redak  $p$  koji odgovara izlaznom vektoru

$$p = \operatorname{argmin}_{i \in \{1,2\}} \{ x_{[i]} / B^{-1} A_{i1} ; B^{-1} A_{i1} > 0 \} = \operatorname{argmin} \{ 3/2, 4/1 \} = 1$$

Pivot (1,1): u bazu ulazi  $a_1$ , a izlazi  $a_3$

Gauss-Jordanova eliminacija tako da  $a_1 = [0, 1, 0]^T$



# Simplex – primjer

Tablični zapis je

	$a_1$	$a_2$	$a_3$	$a_4$	RHS	
$c^T$	-7	-6	0	0	0	$= r^T$
	2	1	1	0	3	
	1	4	0	1	4	

- Pivot (1,1): u bazu ulazi  $a_1$ , a izlazi  $a_3$

	$a_1$	$a_2$	$a_3$	$a_4$	RHS
$r^T$	0	-5/2	7/2	0	21/2
	1	1/2	1/2	0	3/2
	0	7/2	-1/2	1	5/2

Nova baza  $B_1 = [a_1 \ a_4] = I_2$

- rješenje  $x_{(1)} = [3/2 \ 0 \ 0 \ 5/2]^T$ ,  $f(x_{(1)}) = -21/2$



# Simplex – primjer

	$a_1$	$a_2$	$a_3$	$a_4$	RHS
$r^T$	0	-5/2	7/2	0	21/2
	1	1/2	1/2	0	3/2
	0	7/2	-1/2	1	5/2

- Negativan faktor redukcije  $r_2$  - nije optimum!
- Odabir stupca  $q=2$
- Odabir retka
  - $p = \operatorname{argmin}_{i \in \{1,2\}} \{ x_{[i]} / B^{-1}A_{i2} ; B^{-1}A_{i2} > 0 \} = \operatorname{argmin}\{3, 5/7\} = 2$
- Pivot (2,2)



# Simplex – primjer

	$a_1$	$a_2$	$a_3$	$a_4$	RHS
$r^T$	0	-5/2	7/2	0	21/2
	1	1/2	1/2	0	3/2
	0	7/2	-1/2	1	5/2

- Pivot (2,2): u bazu ulazi  $a_2$ , a izlazi  $a_4$

	$a_1$	$a_2$	$a_3$	$a_4$	RHS
$r^T$	0	0	22/7	5/7	86/7
	1	0	4/7	-1/7	8/7
	0	1	-1/7	2/7	5/7



# Simplex – primjer

	$a_1$	$a_2$	$a_3$	$a_4$	RHS
$r^T$	0	0	$22/7$	$5/7$	$86/7$
	1	0	$4/7$	$-1/7$	$8/7$
	0	1	$-1/7$	$2/7$	$5/7$

- nema negativnih faktora redukcije
  - Optimum!
- Baza  $B_2 = [a_1 \ a_2] = I_2$ 
  - Rješenje  $x^* = [8/7 \ 5/7 \ 0 \ 0]^T$ ,  $f(x^*) = -86/7$
- Rješenje polaznog problema
  - $x_1 = 8/7$  i  $x_2 = 5/7$
  - $f_{\max} = 86/7$



# Simplex – problem početne baze!

- Nekad se nakon pretvorbe u standardni oblik ne vidi odmah bazično rješenje!
- Dvofazni simpleks – rješavaju se 2 LPa u nizu
  - 1. FAZA – **pomoćni umjetni korak** za naći početno bazično rješenje
    - Uvijek ima svoju trivijalnu početnu bazu (tako je konstruiran)
  - 2. FAZA – zapravo riješava LP od interesa



# Dvofazni simpleks – umjetni problem

- Pretp. da imamo problem u standardnoj formi

minimizirati  $c^T x$

uz uvjet  $Ax = b$

$x \geq 0$

pri čemu imamo  $m$  ograničenja jednakosti. Dodajemo  $m$  umjetnih varijabli da bismo stvorili jediničnu podmatricu

**NOVI PROBLEM LP':**

minimizirati  $1^T x_{n+1:n+m}$

uz uvjet  $[A \mid I_m] [x_{1:n} \mid x_{n+1:n+m}] = b$

$x_{1:n+m} \geq 0$



# Dvofazni simpleks – 1. FAZA

- Riješimo novi problem već definiranim postupkom
- Tri moguća ishoda:
  1. Optimum  $f_{LP}^* \neq 0 \Rightarrow$  originalni LP neizvediv! **KRAJ!**
  2. Optimum  $f_{LP}^* = 0$ 
    - a) Sve umjetne varijable su nebazične. Adaptacija za 2. FAZU
    - b) Neke umjetne varijable su bazične. Izvodi iteracije simpleksa dok ne izađu sve umjetne varijable iz baze. Adaptacija za 2. FAZU



# Dvofazni simpleks – 2. FAZA

- Adaptacija tablice od LP' (sadrži bazu za originalni LP)
  1. Pobrisati kolone umjetnih varijabli
  2. Zamijeniti fju cilja originalnom
  3. Dovesti prvi red u faktore redukcije
- Riješiti LP od te točke nadalje



# Dvofazni simpleks – primjer

$$\begin{array}{ll}\text{min} & 2x_1 + 3x_2 \\ \text{uz} & 4x_1 + 2x_2 \geq 12 \\ & x_1 + 4x_2 \geq 6 \\ & x_1, x_2 \geq 0\end{array}$$

standardna forma:

$$\begin{array}{lll}\text{min} & 2x_1 + 3x_2 \\ \text{uz} & 4x_1 + 2x_2 - x_3 & = 12 \\ & x_1 + 4x_2 & - x_4 = 6 \\ & x_1, x_2, x_3, x_4 \geq 0\end{array}$$



# Dvofazni simpleks – primjer

Dodajemo dvije nove varijable i novu fju cilja

	$a_1$	$a_2$	$a_3$	$a_4$	$a_5$	$a_6$	<b>b = RHS</b>
$c^T$	0	0	0	0	1	1	0
	4	2	-1	0	1	0	12
	1	4	0	-1	0	1	6



# Dvofazni simpleks – primjer

*Nakon nekoliko iteracija...*

	$a_1$	$a_2$	$a_3$	$a_4$	$a_5$	$a_6$	RHS
$r^T$	0	0	0	0	1	1	0
	1	0	$-2/7$	$1/7$	$2/7$	$-1/7$	$18/7$
	0	1	$1/14$	$-2/7$	$-1/14$	$2/7$	$6/7$

baza  $\mathbf{B}_2 = [\mathbf{a}_1 \ \mathbf{a}_2] = \mathbf{I}_2$

faktori redukcije su nenegativni -> **OPTIMUM** umjetnog problema

umjetne varijable su =0 i umjetna ciljna funkcija =0

gotova 1. FAZA



# Dvofazni simpleks – primjer

2. faza – iz prethodne tablice se izbace stupci umjetnih varijabli i zamijeni se ciljna fja

	$a_1$	$a_2$	$a_3$	$a_4$	RHS
$\mathbf{c}^T$	2	3	0	0	0
	1	0	$-2/7$	$1/7$	$18/7$
	0	1	$1/14$	$-2/7$	$6/7$

**Ispravak 0-tog retka - eliminacijom iznad baznih stupaca**



# Dvofazni simpleks – primjer

	$a_1$	$a_2$	$a_3$	$a_4$	RHS
$r^T$	0	0	5/14	4/7	-54/7
	1	0	-2/7	1/7	18/7
	0	1	1/14	-2/7	6/7

- nema negativnih faktora redukcije
  - **OPTIMUM**
  - Rješenje proširenog izvornog problema je  
 $x = [18/7 \ 6/7 \ 0 \ 0]^T$
  - rješenje izvornog problema  $x = [18/7 \ 6/7]^T$ ,  $f(x)=54/7$



# Dualnost

- Teorija nastala poopćenjem metode Lagrangeovih množitelja
- Svaki LP (kojeg ćemo zvati „primal“) ima svoj povezani LP kojeg zovemo „dual“



# Dualnost – kanonska forma

minimizirati       $c^T x$   
uz uvjet             $Ax \leq b$   
                         $x \geq 0$

- Dual:

maksimizirati       $b^T y$   
uz uvjet             $A^T y \geq c$   
                         $y \geq 0$



# Dual - izvođenje

primal	dual
broj ograničenja	broj varijabli
broj varijabli	broj ograničenja
rhs	funkcija cilja
funkcija cilja	rhs
$A$ matrica koeficijenata	$A^T$
jednakost	urs varijabla
urs varijabla	jednakost
$\leq$ ograničenje	$\geq$ varijabla
$\geq$ ograničenje	$\leq$ varijabla
$\geq$ varijabla	$\geq$ ograničenje
$\leq$ varijabla	$\leq$ ograničenje



# Veze duala i primala - teoremi

- „Dual duala je primal”
- Slaba dualnost
- Jaka dualnost
- Komplementarnost



# Veze duala i primala - teoremi

- Slaba dualnost
  - Ako je  $x$  izvedivo primalno rješenje i  $y$  je izvedivo dualno rješenje, tada je  $y^T b \leq c^T x$
- Jaka dualnost
  - Ako linearni program ima optimalno rješenje, onda ga ima i dual i njihove vrijednosti su jednake.



# Veze duala i primala - teoremi

## • Komplementarnost

- Ako su  $x$  i  $y$  izvediva rješenja primala i duala, onda su ***optimalna ako i samo ako*** vrijedi:

$$x_j(c_j - y^T A_{:,j}) = 0, \forall j$$

Faktor  
redukcije  
varijable  $x_j$

$$y_i(A_{i,:}x - b_i) = 0, \forall i$$

Dopunjene  
i-tog  
ograničenja  
primala



# Dualnost - korisnost

- Ekonomski interpretacija – cijene nad ograničenim resursima
- Minimax teorem u teoriji igara
- Analiza osjetljivosti
- Dualna simpleksna metoda

• ...

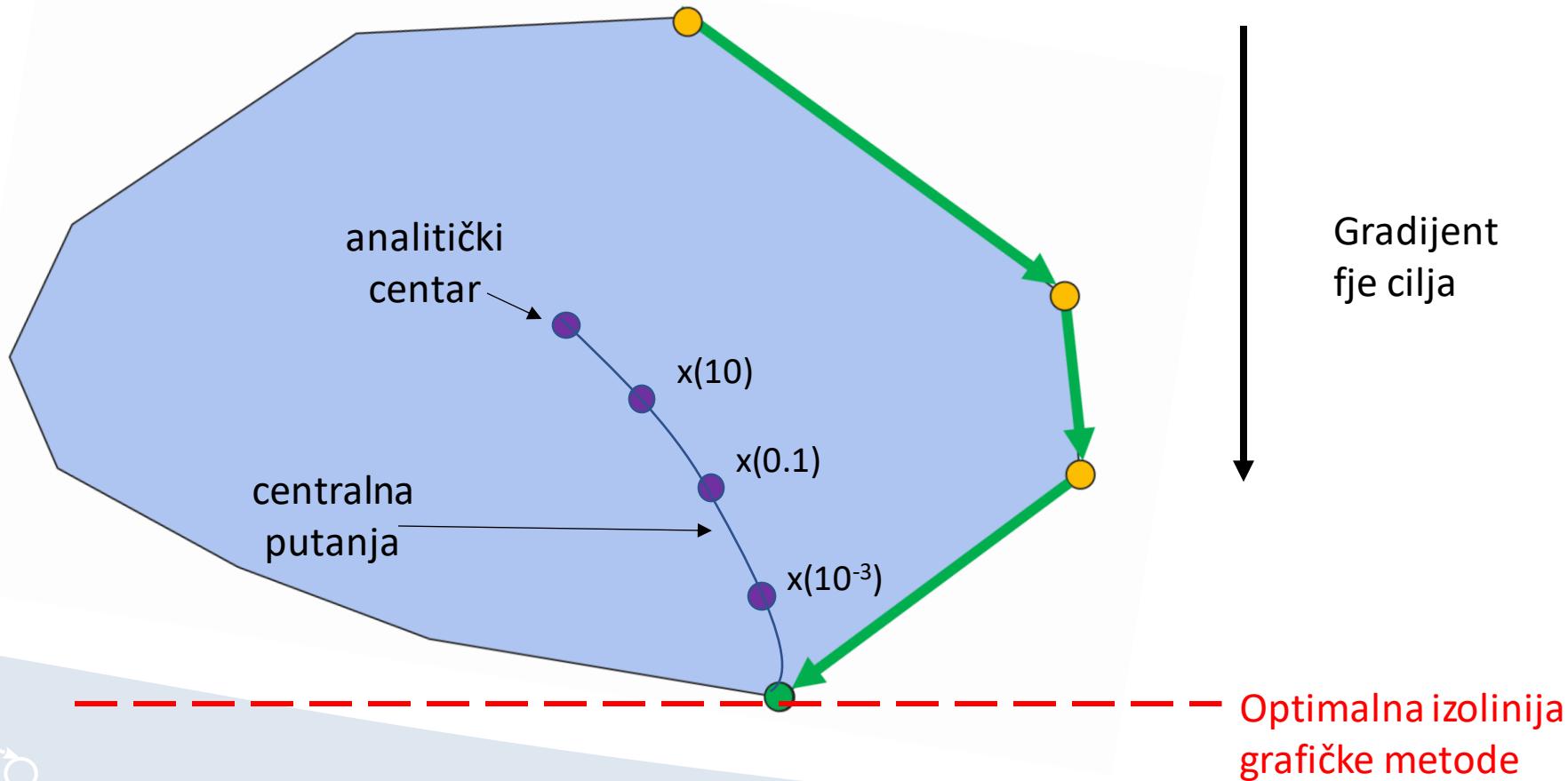


# Simplex – problem!

- Klee-Minty 1972. – konstrukcija perturbirane jedinične hiperkocke za popularna pravila biranja pivota
- Simplex u najgorem slučaju ima eksponencijalnu složenost
- LP je unutar klase problema P



# Metoda unutarnjih točaka - ideja



# Metoda unutarnjih točaka – ideja 1/3

$$\begin{array}{ll} \min & \mathbf{c}^T \mathbf{x} \\ \text{uz uvjet} & \mathbf{Ax} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array}$$

---

$$\begin{array}{ll} \max & \mathbf{b}^T \mathbf{y} \\ \text{uz uvjet} & \mathbf{A}^T \mathbf{y} + \mathbf{s} = \mathbf{c} \\ & \mathbf{s} \geq \mathbf{0} \end{array}$$



# Metoda unutarnjih točaka – ideja 2/3

min  
uz uvjet

$$\begin{aligned} & \mathbf{c}^T \mathbf{x} - \mu \mathbf{1}^T \log(\mathbf{x}) \\ & \mathbf{A} \mathbf{x} = \mathbf{b} \end{aligned}$$

Barijerni  
problemi!

max  
uz uvjet

$$\begin{aligned} & \mathbf{b}^T \mathbf{y} + \mu \mathbf{1}^T \log(\mathbf{s}) \\ & \mathbf{A}^T \mathbf{y} + \mathbf{s} = \mathbf{c} \end{aligned}$$

Logaritamska  
barijera



# Metoda unutarnjih točaka – ideja 3/3

## KKT uvjeti za $\mu$ -barijerne probleme

$$Ax(\mu) = b$$

$$x(\mu) \geq 0$$

$$A^T y(\mu) + s(\mu) = c$$

$$s(\mu) \geq 0$$

$$x(\mu)S(\mu)^{-1} = 1\mu$$

pri čemu  $X(\mu) = \text{diag}(x(\mu))$ ,  $S(\mu) = \text{diag}(s(\mu))$



# Primalni algoritam praćenja putanje

- Barijerni problem „pretežak“ iz KKT
- Taylorov raspis barijerne fje cilja do kvadratnog člana
- Optimizacija Taylorove aproksimacije metodom Lagrangeovih množitelja za pronađazak minimizirajućeg smjera iz trenutne točke



# 06 – Dynamic Programming

*Advanced Algorithms and Data Structures*



UNIVERSITY OF ZAGREB  
Faculty of Electrical  
Engineering and  
Computing

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# Fibonacci numbers (1)

- $F_1 = F_2 = 1, F_n = F_{n-1} + F_{n+2}$
- How many calculations do we make?
- How many are repeated?
- Tree of all calculation cases

```
def fib(n):  
    if n == 1 or n == 2:  
        return 1  
    else:  
        return fib(n-1) + fib(n-2)
```

# Fibonacci numbers (2)

- Time complexity:

$$T_n = T_{n-1} + T_{n-2} + O(1) = \varphi^n$$

$$T_n \geq 2T_{n-1} \Rightarrow \text{at least } O(2^{\frac{n}{2}})$$

```
def fib(n):  
    if n == 1 or n == 2:  
        return 1  
    else:  
        return fib(n-1) + fib(n-2)
```

# Dynamic programming (1)

- Strategy/paradigm for algorithm design
- Method in mathematics
- “Programming” – refers to strategy
- Bellman’s optimality principle

# Dynamic programming (2)

- Bottom-up approach: combine a solution from less complex solutions
  - Unlike divide and conquer, which is top-down
- Applicable when there are **overlapping subproblems**
- Major idea – **reuse calculation results**
- **Memoization** - method of tabularly storing calculation results

# Dynamic programming (3)

- Used for problems such as:
  - One-dimensional distribution
  - Two-dimensional distribution
  - Shortest path
  - Dynamics of equipment replacement
- Or to solve Advent of Code problems:  
<https://adventofcode.com/2023/day/12>

# DP problem features

- **Optimal substructure**
  - Each problem's optimal solution contains the **optimal** solutions of **independent\*** subproblems
  - Form a DAG
  - Not enough -> points to a greedy algorithm; e.g. Dijkstra
- **Overlapping subproblems**
  - The problem's solution requires (leads to) multiple solutions to **identical subproblems**

\*Subproblems are independent regarding one another (when on the same level)

# Solving a DP problem (1)

4 steps:

## 1. Characterize the structure of an optimal solution

- a) Does the problem meet the requirements for an optimal solution?
- b) Are the problems overlapping?
  - Solvability of the problem via DP – intuition

## 2. Recursively define the value of an optimal solution

- “value” is from the function being optimized

# Solving a DP problem (1)

## 3. Compute the value of an optimal solution

- Using bottom-up and memoization

## 4. Construct an optimal solution from the computed information

- From here on – the problem is solved
- Packs the entire solution as a set of decisions
- If we just need a value, then this is skipped

# Fibonacci numbers – DP approach

- Implement memorization
- Dictionary to store calculated results
- Still recursive

```
def dyn_fib(n, memo: dict):  
    if n == 1 or n == 2:  
        memo[n] = 1  
        return 1  
  
    else:  
        if n-1 not in memo:  
            dyn_fib(n-1, memo)  
        if n-2 not in memo:  
            dyn_fib(n-2, memo)  
        memo[n] = memo[n-1] + memo[n-2]  
    return memo[n]
```

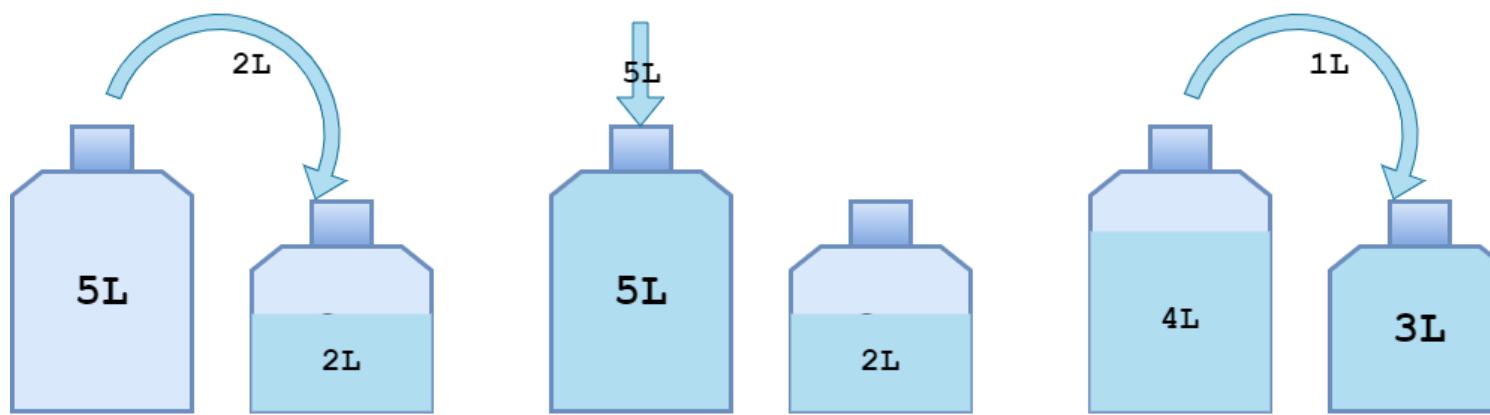
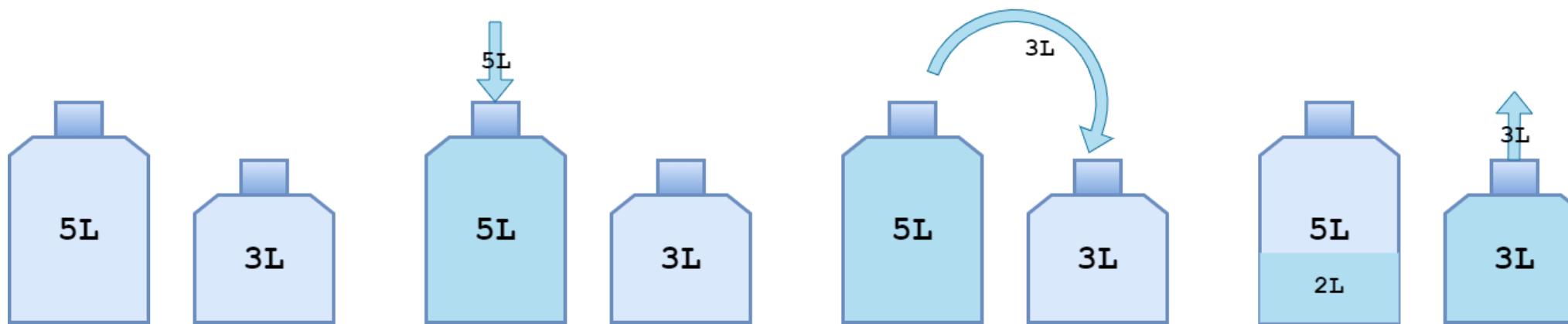
# Example 2 - Factorials

Calculate the factorial  $n!$ .

- $n! = 1 \cdot 2 \cdot 3 \cdot \dots \cdot (n - 1) \cdot n$
- We just need a value, not the decisions.
- What are the subproblems in relation to the problem?
  - Do they overlap?
- Can we define the solution recursively?
- Can we use memoization for computation?
- At which step do we stop?
  - Do we care about the decisions?

# Example 3 – Jugs

- Problem from “Die Hard 2”
- Fill any of the jugs with 4L



Is the value 4L enough?

# Knapsack problem (1)

Given a set of items, each with a **cost** (weight) and a **value**, determine which items to include in the collection so that the **total cost is less than or equal to a given limit** and the **total value is as large as possible**.



How many DP steps are we considering?

# Knapsack problem (2)

Types of knapsack problems:

- **0-1 knapsack problem**
  - Finite set of items  $i$ , having one of each kind
  - We use “0” and “1” to denote that an item doesn’t exist ( $x_i$ )
- **Bounded knapsack problem**
  - Finite set of items, but each item can have up to  $c$  copies
- **Unbounded knapsack problem**
  - No upper bounds for each item

# Knapsack problem (3)

Other types of knapsack problems:

- Multi-dimensional knapsack problem
- Multiple knapsack problem
- Quadratic knapsack problem

# Knapsack problem – formalized (1)

## 0-1 knapsack problem

$$\text{maximize} \quad \sum_{i=1}^n x_i v_i$$

$$\text{subject to} \quad \sum_{i=1}^n x_i w_i \leq W$$
$$x_i \in \{0, 1\}$$

$x_i$	$i^{\text{th}}$ item selection
$w_i$ (or $c_i$ )	$i^{\text{th}}$ item weight or cost
$v_i$	$i^{\text{th}}$ item value
$W$	Maximum weight or cost

# Knapsack problem – formalized (2)

## Bounded knapsack problem

$$\text{maximize} \quad \sum_{i=1}^n x_i v_i$$

$$\text{subject to} \quad \sum_{i=1}^n x_i w_i \leq W$$
$$x_i \in [0, c]$$

$x_i$	$i^{\text{th}}$ item selection
$w_i$ (or $c_i$ )	$i^{\text{th}}$ item weight or cost
$v_i$	$i^{\text{th}}$ item value
$W$	Maximum weight or cost
$c$	Maximum copies of each item

# Knapsack problem – formalized (3)

## Unbounded knapsack problem

$$\text{maximize} \quad \sum_{i=1}^n x_i v_i$$

$$\text{subject to} \quad \sum_{i=1}^n x_i w_i \leq W$$

$$x_i \in \mathbb{N}$$

$x_i$	i <sup>th</sup> item selection
$w_i$ (or $c_i$ )	i <sup>th</sup> item weight or cost
$v_i$	i <sup>th</sup> item value
$W$	Maximum weight or cost

# Knapsack – a practical example (1)

- Studying for an exam on a tight schedule
- Topics
  - Time required to learn
  - The number of points it will bring on the exam

	<b>Topic 1</b>	<b>Topic 2</b>	<b>Topic 3</b>	<b>Topic 4</b>
<b>Points</b>	10	8	15	7
<b>Cost in time</b>	5	2	7	3

$$c_{max} = 12 \text{ (hours)}$$

# Knapsack – example (2)

## Step 1: Characterize the structure of an optimal solution

- We assume the optimal solution has  $k$  topics and the entire time is used
- The optimal chosen topics set:  $\Omega$
- If we remove any topic, then we have  $\Omega$  for that capacity and  $k - 1$  topics
- The optimal solution is reached through the optimal solutions on smaller sets of chosen topics – **optimal substructure**
- The solution for  $k$  topics is reached by expanding from the solution for  $k - 1$  topics, then that on  $k - 2$  topics... – a **overlapping problems**

# Knapsack – example (3)

## Step 2: Recursively define the value of an optimal solution

- We iteratively decide if we want to include a topic at some available amount of time

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - cost(k)] + value(k) \}$$

$$v_0(\cdot) = 0$$

# Knapsack – example (4)

## Step 3: Compute the value of an optimal solution

- Create a table with expected points depending on the time capacity

# Knapsack – example (5)

Cap	T1	T2	T3	T4
1	0			
2	0			
3	0			
4	0			
5	10			
6				
7				
8				
9				
10				
11				
12				

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - cost(k)] + value(k) \}$$

# Knapsack – example (6)

Cap	T1	T2	T3	T4
1	0	0		
2	0	8		
3	0			
4	0			
5	10			
6	10			
7	10			
8	10			
9	10			
10	10			
11	10			
12	10			

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - cost(k)] + value(k) \}$$

We now observe both T1 and T2 values

# Knapsack – example (7)

Cap	T1	T2	T3	T4
1	0	0		
2	0	8		
3	0	8		
4	0	8		
5	10	10		
6	10	10		
7	10			
8	10			
9	10			
10	10			
11	10			
12	10			

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - cost(k)] + value(k) \}$$

# Knapsack – example (8)

Cap	T1	T2	T3	T4
1	0	0		
2	0	8		
3	0	8		
4	0	8		
5	10	10		
6	10	10		
7	10	18		
8	10	18		
9	10	18		
10	10	18		
11	10	18		
12	10	18		

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - cost(k)] + value(k) \}$$

# Knapsack – example (9)

Cap	T1	T2	T3	T4
1	0	0	0	
2	0	8	8	
3	0	8	8	
4	0	8	8	
5	10	10	10	
6	10	10	10	
7	10	18	18	
8	10	18	18	
9	10	18		
10	10	18		
11	10	18		
12	10	18		

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - \text{cost}(k)] + \text{value}(k) \}$$

15 < 18 – keeping T1 and T2

# Knapsack – example (10)

Cap	T1	T2	T3	T4
1	0	0	0	
2	0	8	8	
3	0	8	8	
4	0	8	8	
5	10	10	10	
6	10	10	10	
7	10	18	18	
8	10	18	18	
9	10	18	23	
10	10	18	23	
11	10	18	23	
12	10	18		

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - cost(k)] + value(k) \}$$

8+15 > 18 – keeping T2 and T3

# Knapsack – example (11)

Cap	T1	T2	T3	T4
1	0	0	0	
2	0	8	8	
3	0	8	8	
4	0	8	8	
5	10	10	10	
6	10	10	10	
7	10	18	18	
8	10	18	18	
9	10	18	23	
10	10	18	23	
11	10	18	23	
12	10	18	25	

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - \text{cost}(k)] + \text{value}(k) \}$$

10+15 > 23 – keeping T1 and T3

# Knapsack – example (12)

Cap	T1	T2	T3	T4
1	0	0	0	0
2	0	8	8	8
3	0	8	8	8
4	0	8	8	8
5	10	10	10	15
6	10	10	10	15
7	10	18	18	
8	10	18	18	
9	10	18	23	
10	10	18	23	
11	10	18	23	
12	10	18	25	

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max\{ v_{k-1}(c), v_{k-1}[c - \text{cost}(k)] + \text{value}(k) \}$$

8+7 > 10 – keeping T2 and T4

# Knapsack – example (13)

Cap	T1	T2	T3	T4
1	0	0	0	0
2	0	8	8	8
3	0	8	8	8
4	0	8	8	8
5	10	10	10	15
6	10	10	10	15
7	10	18	18	18
8	10	18	18	18
9	10	18	23	23
10	10	18	23	25
11	10	18	23	25
12	10	18	25	

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max \{ v_{k-1}(c), v_{k-1}[c - \text{cost}(k)] + \text{value}(k) \}$$

T1 + T2 + T4

# Knapsack – example (14)

Cap	T1	T2	T3	T4
1	0	0	0	0
2	0	8	8	8
3	0	8	8	8
4	0	8	8	8
5	10	10	10	15
6	10	10	10	15
7	10	18	18	18
8	10	18	18	18
9	10	18	23	23
10	10	18	23	25
11	10	18	23	25
12	10	18	25	30

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

$$v_k(c) = \max \{ v_{k-1}(c), v_{k-1}[c - \text{cost}(k)] + \text{value}(k) \}$$

T2 + T3 + T4

# Knapsack – example (15)

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

Step 4: Construct an optimal solution from the computed information

- Change in row value right->left signifies an item was added
- Which items – determined by arrows
  - Impl. flag if the column item is included
- $O(C_{max} \cdot N)$

Cap	T1	T2	T3	T4
1	0	0	0	0
2	0	8	8	8
3	0	8	8	8
4	0	8	8	8
5	10	10	10	15
6	10	10	10	15
7	10	18	18	18
8	10	18	18	18
9	10	18	23	23
10	10	18	23	25
11	10	18	23	25
12	10	18	25	30

# Knapsack – human optimization

- Sort by cost
- Ignore capacities < minimum cost

	Topic 1	Topic 2	Topic 3	Topic 4
Value	10	8	15	7
Cost	5	2	7	3

	Topic 2	Topic 4	Topic 1	Topic 3
Value	8	7	10	15
Cost	2	3	5	7

# Knapsack – naïve recursive solution

```
def knapsack(capacity, costs, values, num_items):
    # Base case: If no items are left or capacity is 0
    if num_items == 0 or capacity == 0:
        return 0

    # Check if there is enough capacity for the current item
    if costs[num_items-1] > capacity:
        return knapsack(capacity, costs, values, num_items-1)

    # Return the maximum of two cases:
    # (1) nth item included
    # (2) not included
    else:
        return max(
            knapsack(capacity, costs, values, num_items-1),
            values[num_items-1] + knapsack(capacity-costs[num_items-1], costs,
values, num_items-1)
        )
```

# Knapsack – with memoization

```
def knapsack(capacity, costs, values, num_items, memo={}):
    # Base case: If no items are left or capacity is 0
    if num_items == 0 or capacity == 0:
        return 0

    # Check if the result is already in the memo dictionary
    if (num_items, capacity) in memo:
        return memo[(num_items, capacity)]
    # Capacity check
    if costs[num_items-1] > capacity:
        result = knapsack(capacity, costs, values, num_items-1, memo)
    else:
        # determine maximum
        result = max(
            knapsack(capacity, costs, values, num_items-1, memo),
            values[num_items-1] + knapsack(capacity-costs[num_items-1], costs, values,
num_items-1, memo)
        )
    # Store the result in the memo dictionary
    memo[(num_items, capacity)] = result
    return result
```

# Knapsack – with decisions

```
def knapsack(capacity, costs, values, num_items, memo={}, items_taken=[]):

    # Base case: If no items are left or capacity is 0
    if num_items == 0 or capacity == 0:
        return 0, items_taken

    # Check if the result is already in the memo dictionary
    if (num_items, capacity) in memo:
        return memo[(num_items, capacity)], items_taken

    # determine if the item fits the capacity
    if costs[num_items-1] > capacity:
        result, items_taken = knapsack(capacity, costs, values, num_items-1, memo, items_taken)
    else:
        # calculate the two decision cases
        included_value, included_items = knapsack(capacity-costs[num_items-1], costs, values, num_items-1, memo,
items_taken + [num_items-1])
        excluded_value, excluded_items = knapsack(capacity, costs, values, num_items-1, memo, items_taken)
        # determine MAX
        if included_value + values[num_items-1] > excluded_value:
            result = included_value + values[num_items-1]
            items_taken = included_items
        else:
            result = excluded_value
            items_taken = excluded_items

    # Store the result in the memo dictionary
    memo[(num_items, capacity)] = result
    return result, items_taken
```

# DP example task

- Traverse a matrix field by going down or right-down
- Determine the path with the maximum values passed

5	x	x	x
2	4	x	x
7	9	2	x
7	7	6	7

