



COMP 3200

Artificial Intelligence

Lecture 12

Two-Player Games

Mini-Max Search

Alpha-Beta Pruning

Multi-Player Games

- So far we have only looked at problems with a single agent (single agent search)
- Heuristic search can also be applied to environments (and games) with multiple agents (or players)
- Games can have a number of properties, similar to single agent environments

Fully vs. Partially Observable



Deterministic vs. Stochastic



Dynamic vs. Static



Game Properties: Players

- How many players are in the game?
- Examples of Game Players:
 - Chess, Go: 2 Players
 - Baseball: 18 Players, 2 Teams
 - Starcraft: 1v1, 2v2, 8 Person FFA
 - Poker: 10 People at a table

Game Properties: Payoffs

- What does each player hope to achieve?
- Game Payoff Examples:
 - Prisoner's Dilemma: Maximize Reward
 - Poker: Maximize Profits
 - Chess: Win the game
- Zero Sum Game:
 - Each player's gain or loss of payoff/utility is equally balanced by the utility of the other players
 - Win / Lose games are zero sum

Most Traditional Board Games

- Two-Player
- Zero Sum
- Alternating Move
- Perfect Information
- Deterministic
- Discrete

The Chess-Playing Computer

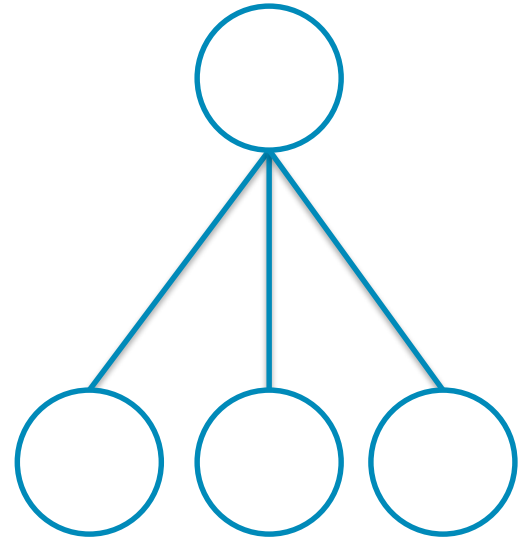


Game-Playing Computer

- How can we design an algorithm to play a two-player alternating move game?
 1. Analysis / Strategy / Tactics
 2. Thousands of If-Then Statements
 3. Look-Ahead and Evaluate

Look-Ahead and Evaluate

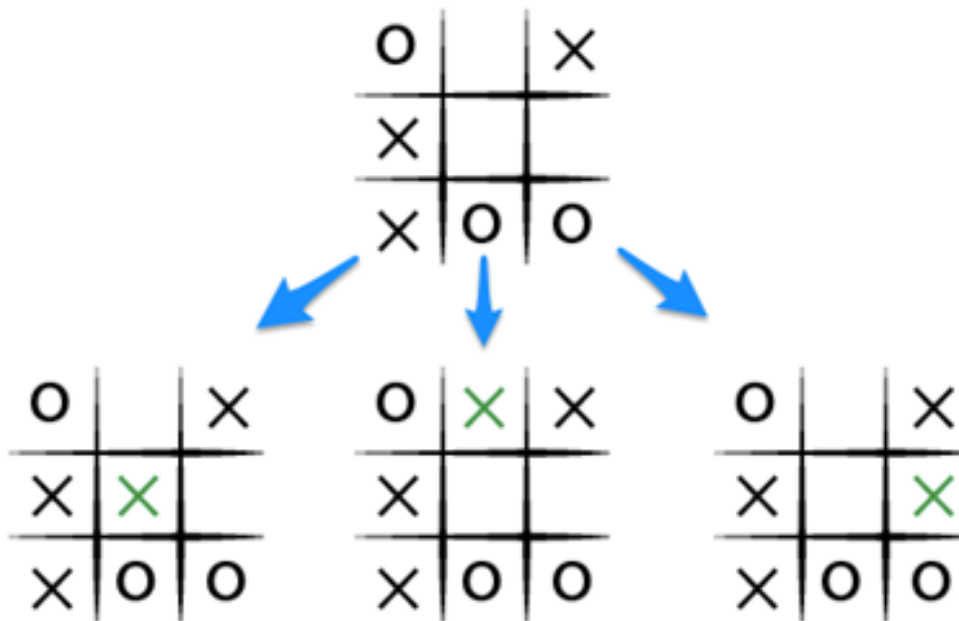
- Generate a list of actions from a given state
- Evaluate those actions based on features of the resulting states
- Do the action which has the highest evaluation



Lookahead and Evaluate

1. Function **LookaheadAndEvaluate**(state)
2. maxVal = -infinity
3. bestAction = null
4. **for** (action : state.getLegalActions())
5. child = state.doAction(action)
6. val = eval(child)
7. **if** (val > maxVal)
8. maxVal = val
9. bestAction = action
10. **return** bestAction

Tic-Tac-Toe Example



Game-Playing Computer

- How can we design an algorithm to play a two-player alternating move game?
 1. Analysis / Strategy / Tactics
 2. Thousands of If-Then Statements
 3. Look-Ahead and Evaluate
 4. Search the Entire Game Tree

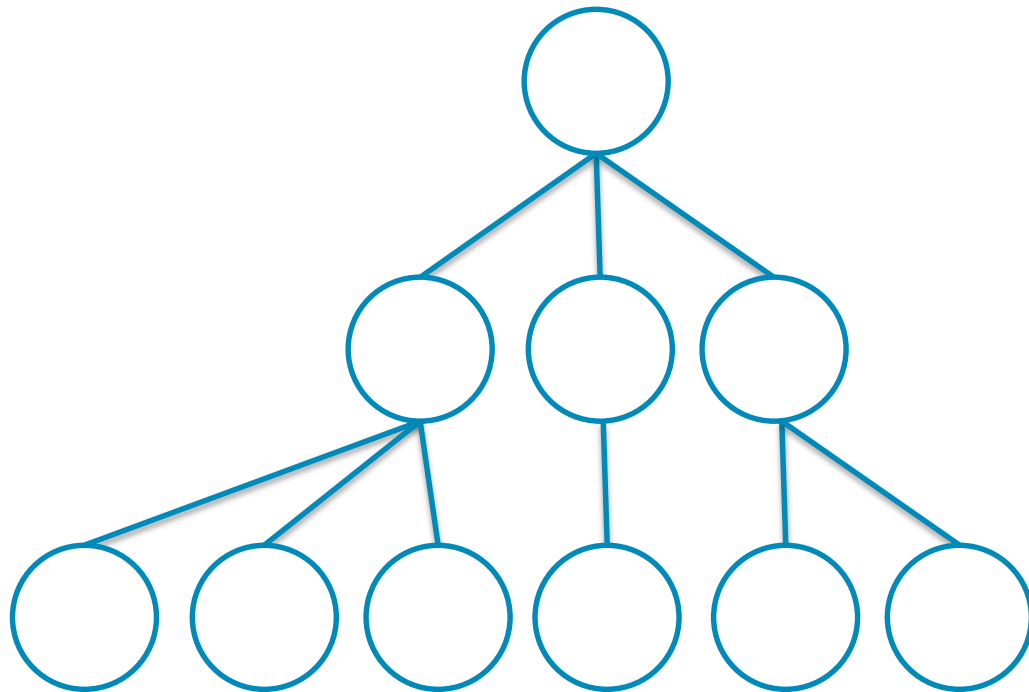
How big is the game tree?

- Some number of actions at each state
 - b = "Branching Factor"
- Game ends after some number of moves
 - d = Search Depth
- Game tree = b^d
- Chess = 10^{120}

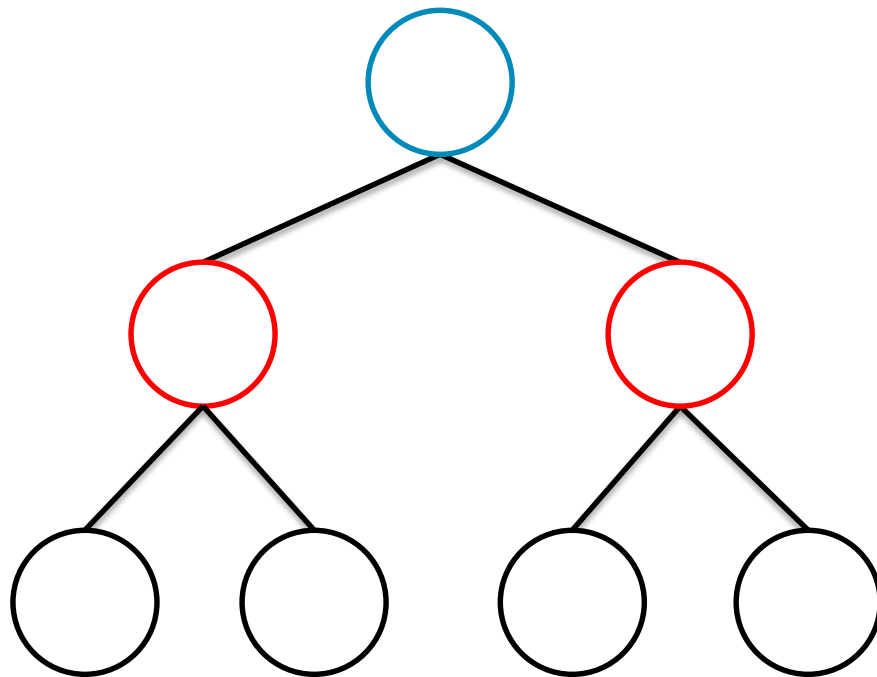
Game-Playing Computer

- How can we design an algorithm to play a two-player alternating move game?
 1. Analysis / Strategy / Tactics
 2. Thousands of If-Then Statements
 3. Look-Ahead and Evaluate
 4. Search the Entire Game Tree
 5. Look-Ahead as Far as Possible

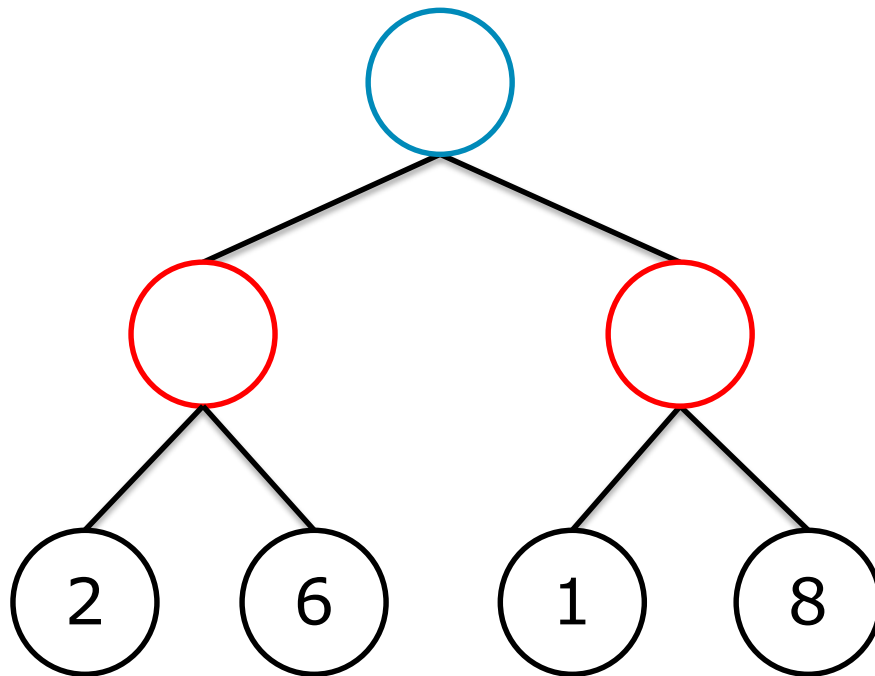
Look Ahead as Far as Possible



Competing Players



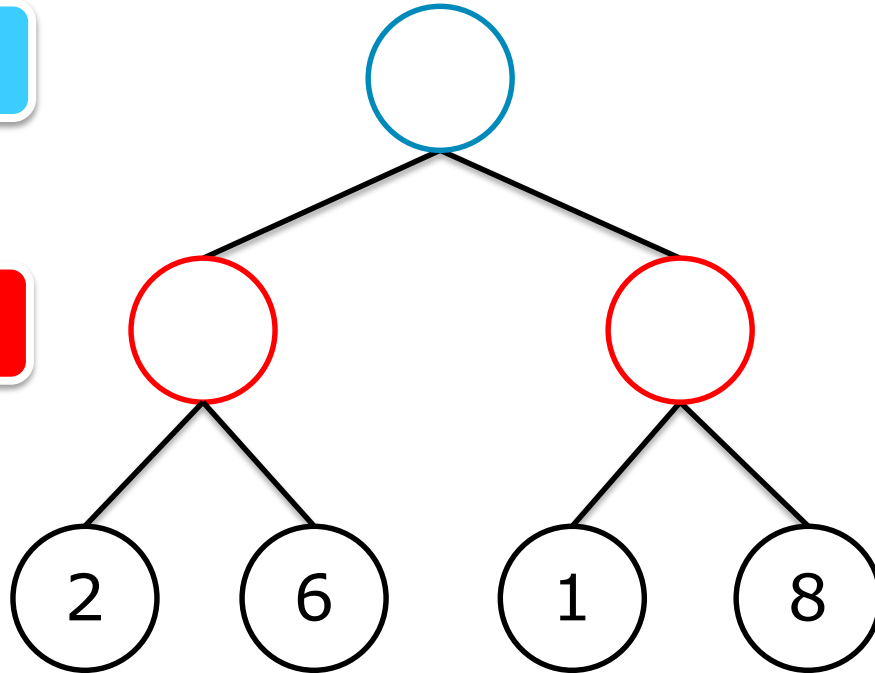
Competing Players



Competing Players

Max Player

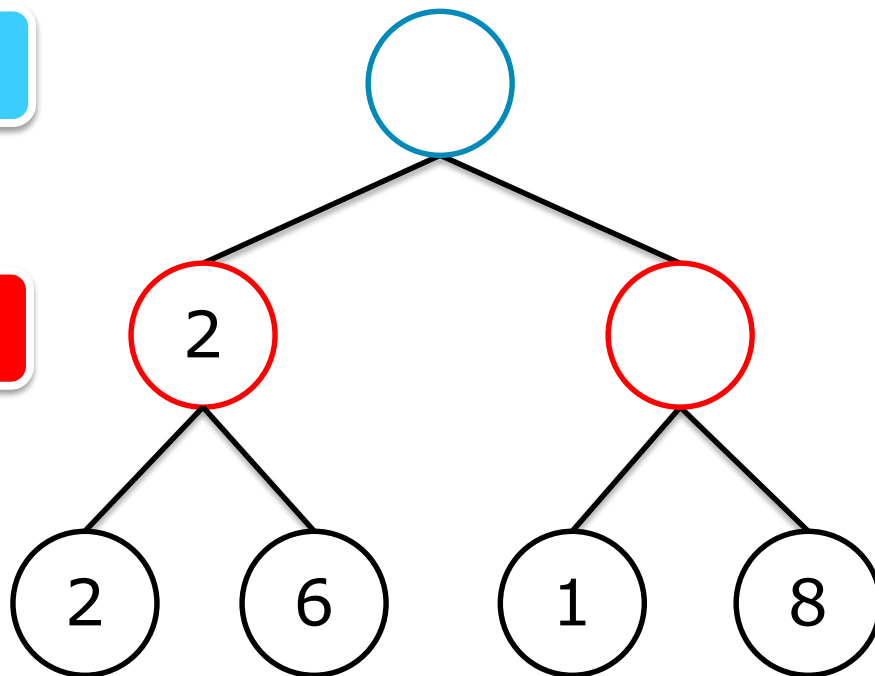
Min Player



Competing Players

Max Player

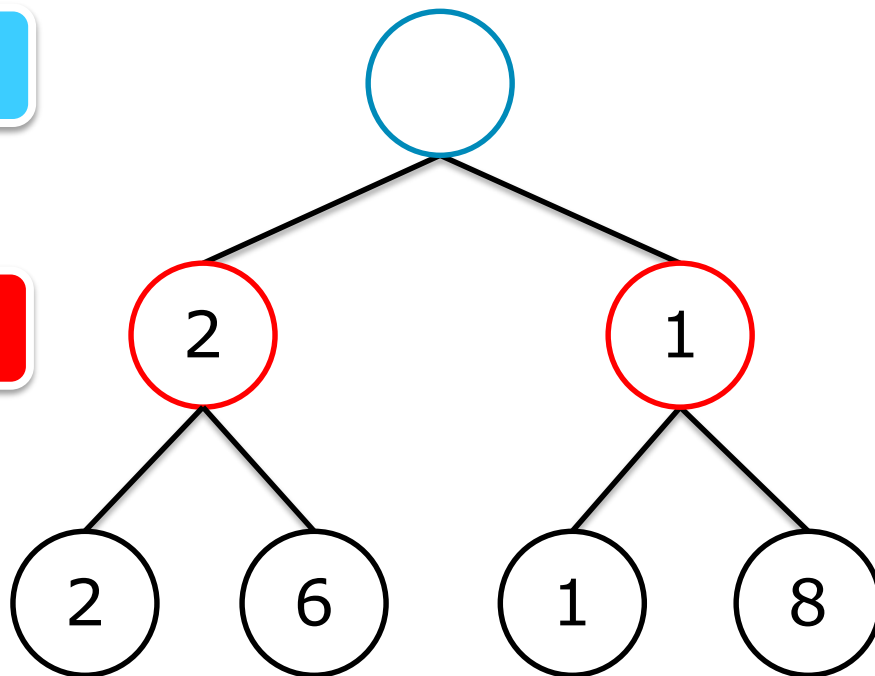
Min Player



Competing Players

Max Player

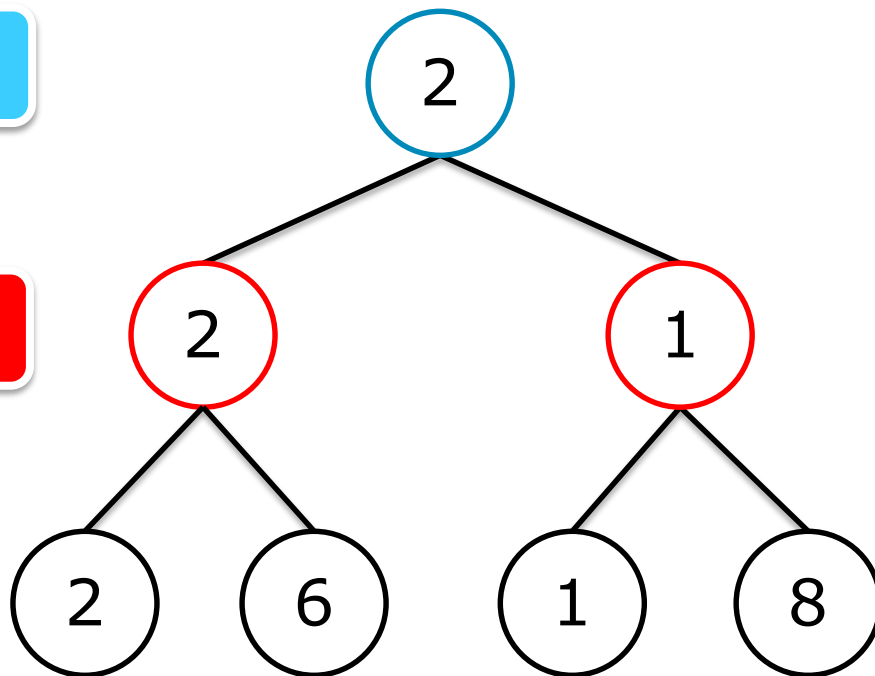
Min Player



Competing Players

Max Player

Min Player



Lookahead and Evaluate

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3. bestAction = null
4. **for** (action : state.getLegalActions())
5. child = state.doAction(action)
6. val = eval(child)
7. **if** (val > maxVal)
8. maxVal = val
9. bestAction = action
10. **return** bestAction

MaxValue (single depth)

1. Function **MaxValue**(s)
2. $v = -\text{infinity}$
3. **for** (c in children(s))
4. $v' = \text{eval}(c)$
5. **if** ($v' > v$) $v = v'$
6. **return** v

MaxValue

1. Function **MaxValue**(s)
2. **if** terminal(s)
3. **return** eval(s)
4. v = -infinity
5. **for** (c in children(s))
6. v' = **MinValue**(c)
7. **if** (v' > v) v = v'
8. **return** v

MaxValue and MinValue (full tree)

```
1. Function MaxValue(s)
2.   if (terminal(s))
3.     return eval(s)
4.   v = -infinity
5.   for (c in children(s))
6.     v' = MinValue(c)
7.     if (v' > v) v = v'
8.   return v
```

```
1. Function MinValue(s)
2.   if (terminal(s))
3.     return eval(s)
4.   v = +infinity
5.   for (c in children(s))
6.     v' = MaxValue(c)
7.     if (v' < v) v = v'
8.   return v
```

eval(s) returns score w.r.t. the maximizing player

MaxValue and MinValue (depth limit)

```
1. Function MaxValue(s, d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   v = -infinity
5.   for (c in children(s))
6.     v' = MinValue(c, d+1)
7.     if (v' > v) v = v'
8.   return v
```

```
1. Function MinValue(s, d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   v = +infinity
5.   for (c in children(s))
6.     v' = MaxValue(c, d+1)
7.     if (v' < v) v = v'
8.   return v
```

eval(s) returns score w.r.t. the maximizing player

Minimax Algorithm

1. Function **MiniMax**(s, d, max)
2. **if** (terminal(s) or d > maxD)
3. **return** eval(s)
4. **if** (max) // maximizing player
5. v = -infinity
6. **for** (c in children(s))
7. v = max(v, **MiniMax**(c, d+1, false))
8. **else** // minimizing player
9. v = +infinity
10. **for** (c in children(s))
11. v = min(v, **MiniMax**(c, d+1, true))

Initial Call: MiniMax(startState, 0, true)

Negamax Algorithm

1. Function **NegaMax**(s, d, player)
2. **if** (terminal(s) d > maxD)
3. **return** eval(s, player)
4. v = -infinity
5. **for** (c in children(s))
6. v = max(v, -**NegaMax**(child, d+1, !player))

Relies on: $\max(a, b) = -\max(-a, -b)$

Initial Call: **NegaMax**(startState, 0, startState.player)

Mini-Max Properties

- Complete and Optimal: Will find the optimal solution to a given max depth
- Each player plays a best response to the possible actions of the other player
- Mini-Max plays the Nash Equilibrium

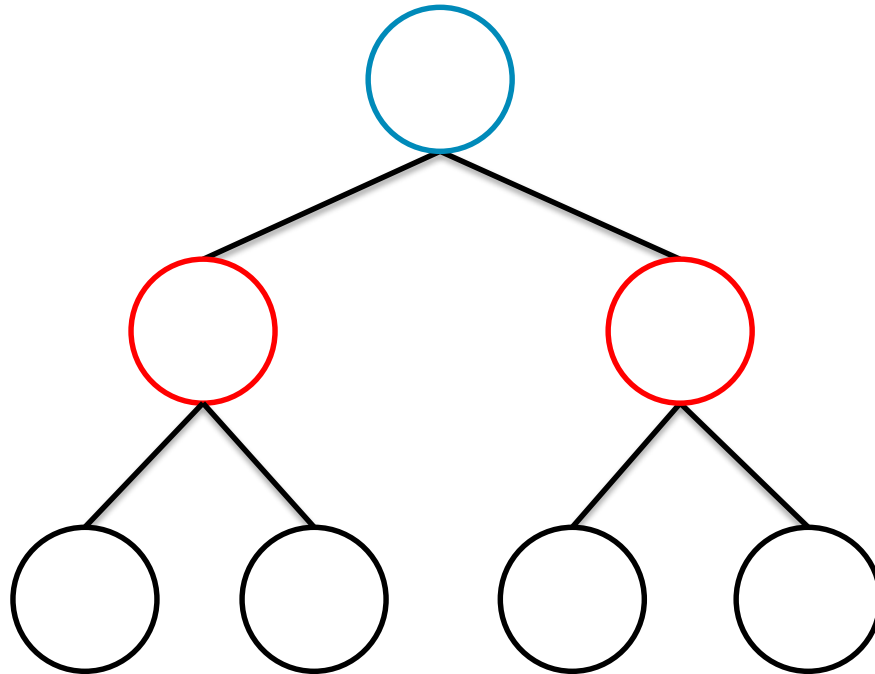
Nash Equilibrium (NE)

- For two-player, finite, zero-sum games, a Nash Equilibrium exists
- Recall that in a NE:
 - Each player is best responding to the other
 - Neither player can gain by deviating
 - Neither player has any regrets
- Playing a NE is a very strong strategy

Improving MiniMax

Max

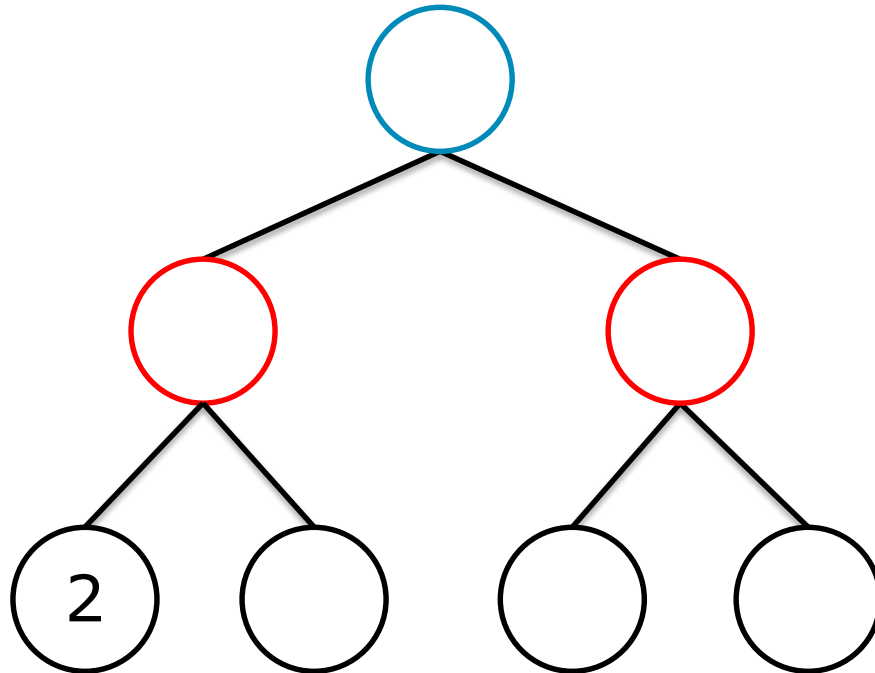
Min



Improving MiniMax

Max

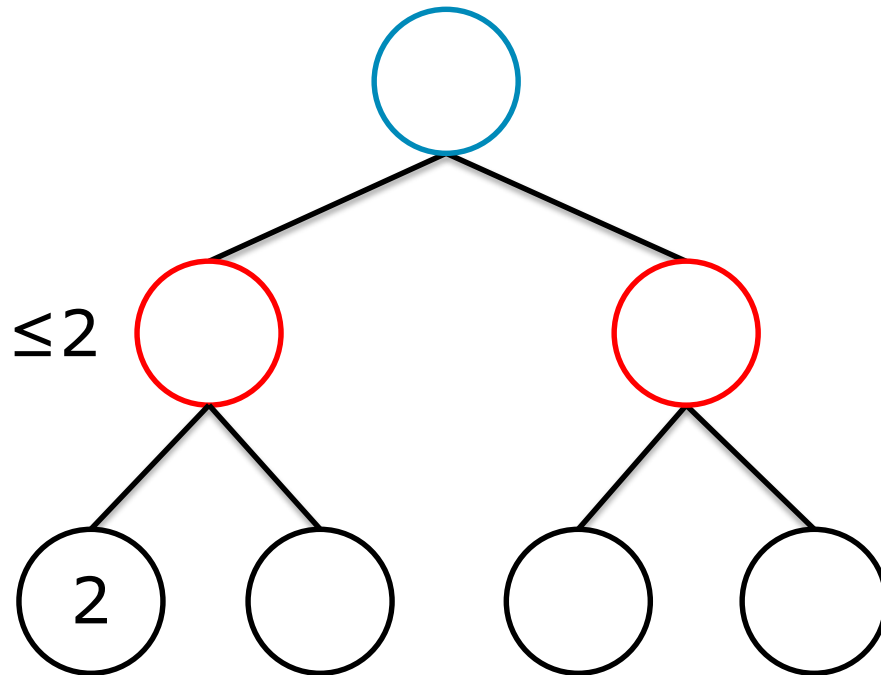
Min



Improving MiniMax

Max

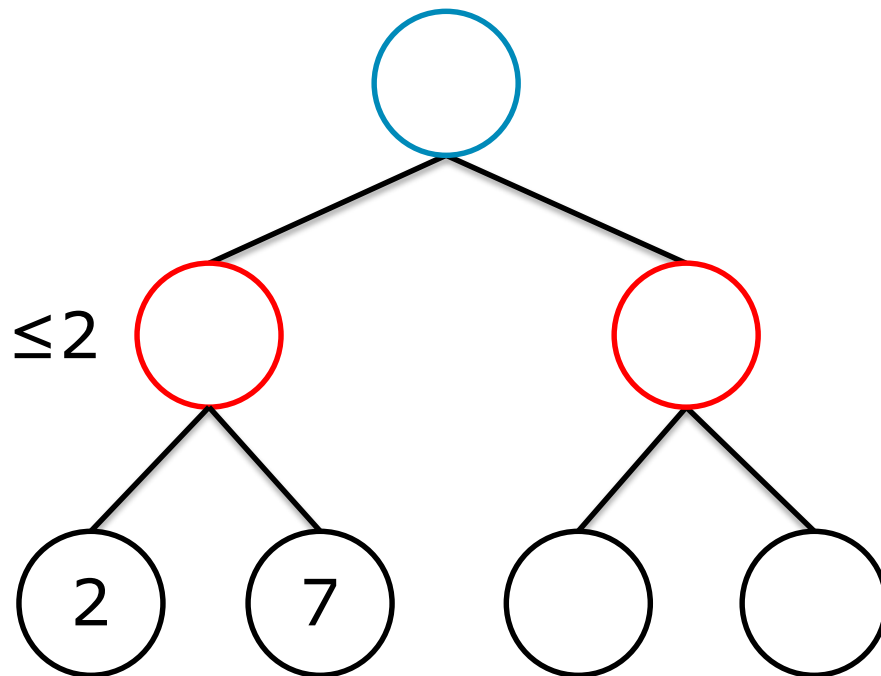
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Improving MiniMax

Max

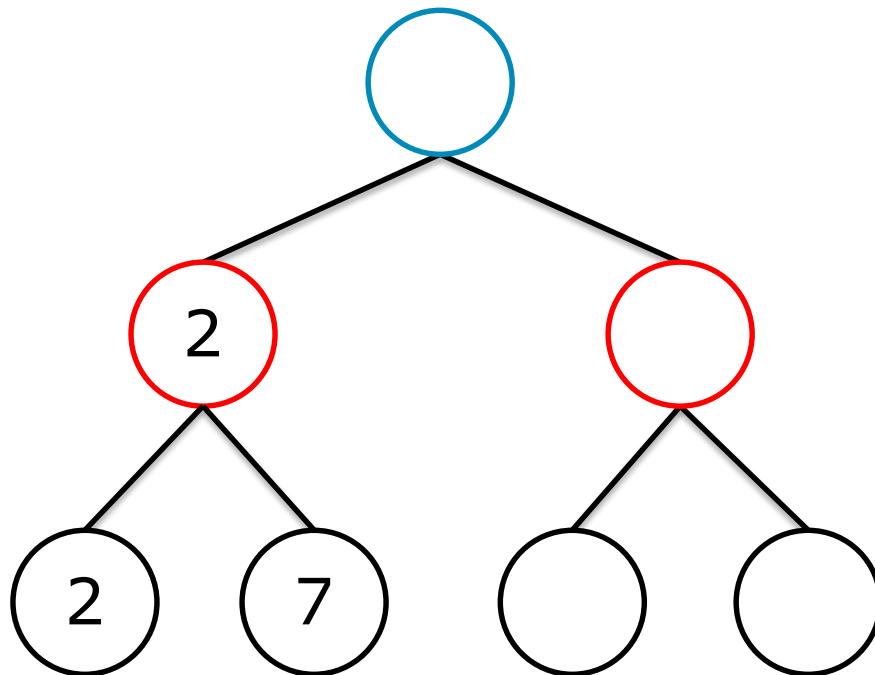
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Improving MiniMax

Max

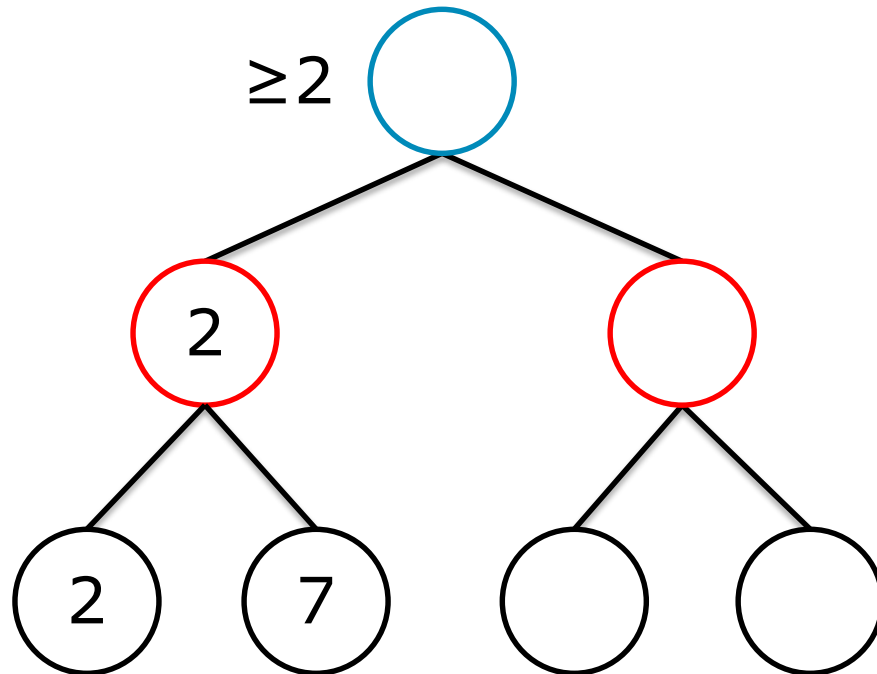
Min



Improving MiniMax

Max

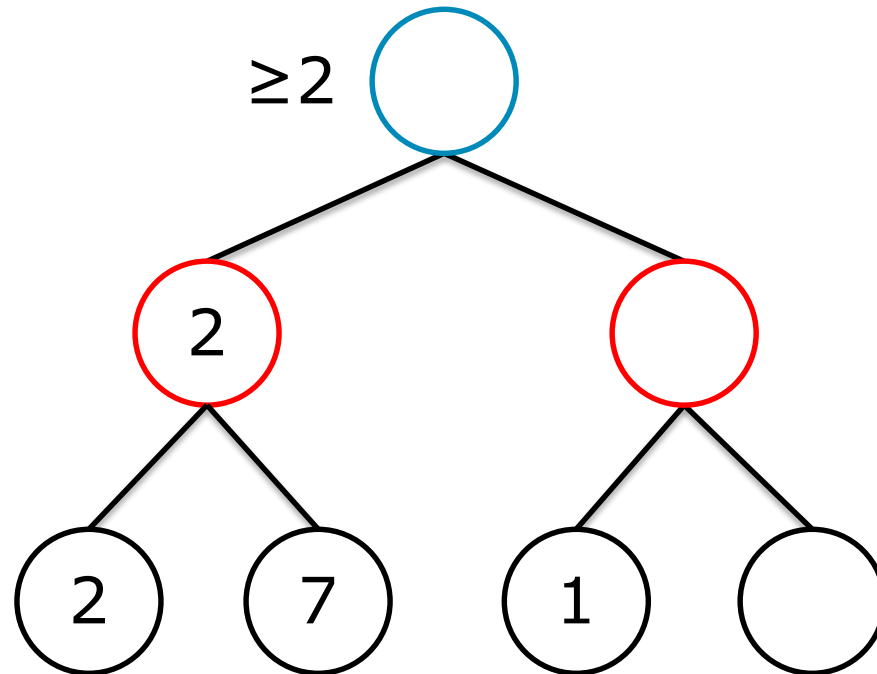
Min



Improving MiniMax

Max

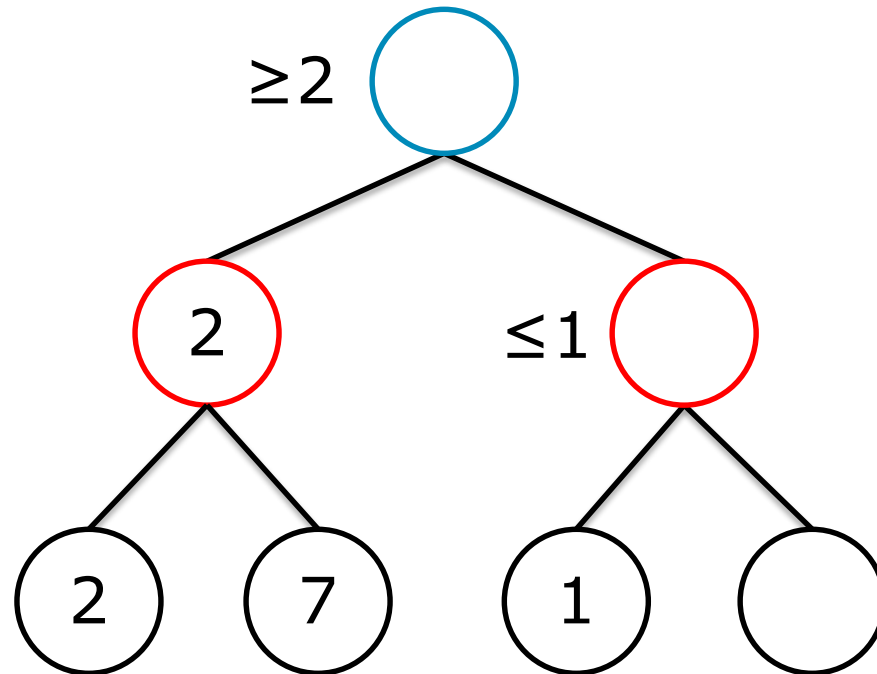
Min



Improving MiniMax

Max

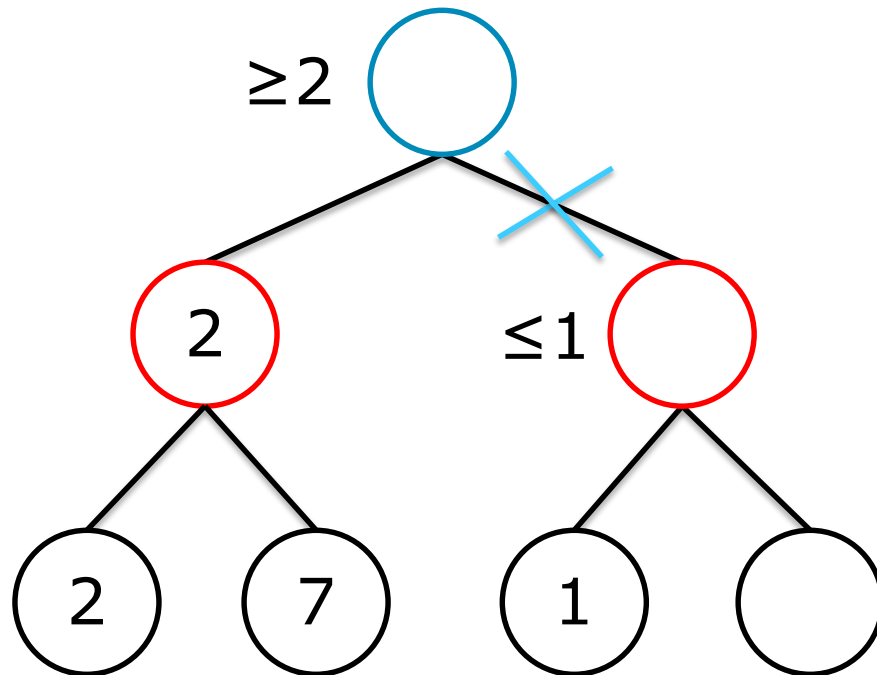
Min



Improving MiniMax

Max

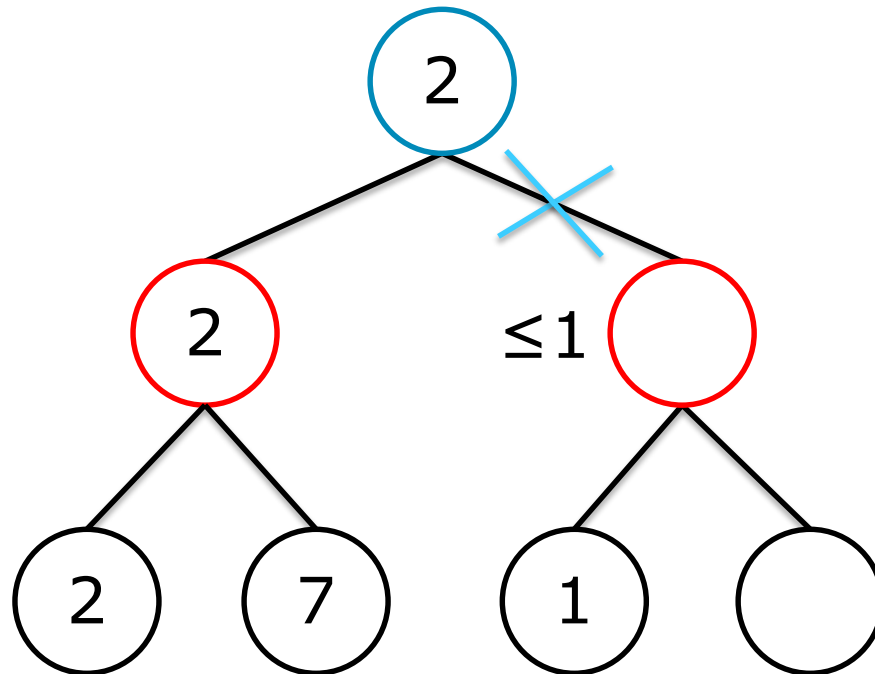
Min



Improving MiniMax

Max

Min

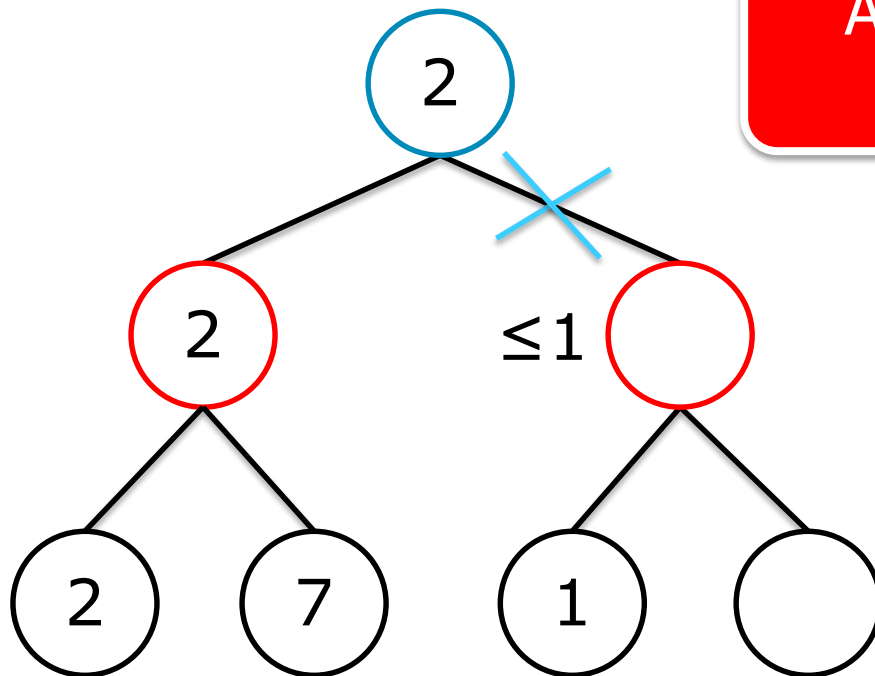


Improving MiniMax

Max

Min

Alpha-Beta
Pruning



Alpha-Beta Pruning

- Alpha-Beta is not a different algorithm than MiniMax, it is an optimization
 - “Minimax with Alpha-Beta Pruning”
- Maintains all of the properties of minimax but is strictly better
- Cutting off branches of the search tree yields exponential savings

Alpha-Beta Pruning

- Uses the MiniMax algorithm
- Introduce 2 new variables
- **Alpha (α)** – Best alternative for Max on this particular path through the tree
- **Beta (β)** – Best alternative for Min on this particular path through the tree
- α , β are an ever narrowing window which 'good' paths pass through, and other paths are pruned

MaxValue and MinValue (minimax)

```
1. Function MaxValue(s, d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   v = -infinity
5.   for (c in children(s))
6.     v' = MinValue(c, d+1)
7.     if (v' > v) v = v'
```

```
8.   return v
```

```
1. Function MinValue(s, d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   v = +infinity
5.   for (c in children(s))
6.     v' = MaxValue(c, d+1)
7.     if (v' < v) v = v'
```

```
8.   return v
```

eval(s) returns score w.r.t. the maximizing player

MaxValue and MinValue (alpha beta)

```
1. Function MaxValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or  $d > \text{maxD}$ )
3.     return eval(s)
4.    $v = -\text{infinity}$ 
5.   for (c in children(s))
6.      $v' = \text{MinValue}(c, \alpha, \beta, d+1)$ 
7.     if ( $v' > v$ )  $v = v'$ 
8.     if ( $v' \geq \beta$ ) return v
9.     if ( $v' > \alpha$ )  $\alpha = v'$ 
10.  return v
```

```
1. Function MinValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or  $d > \text{maxD}$ )
3.     return eval(s)
4.    $v = +\text{infinity}$ 
5.   for (c in children(s))
6.      $v' = \text{MaxValue}(c, \alpha, \beta, d+1)$ 
7.     if ( $v' < v$ )  $v = v'$ 
8.     if ( $v' \leq \alpha$ ) return v
9.     if ( $v' < \beta$ )  $\beta = v'$ 
10.  return v
```

Initial Call: `MaxValue(startState, -infinity, +infinity, 0)`

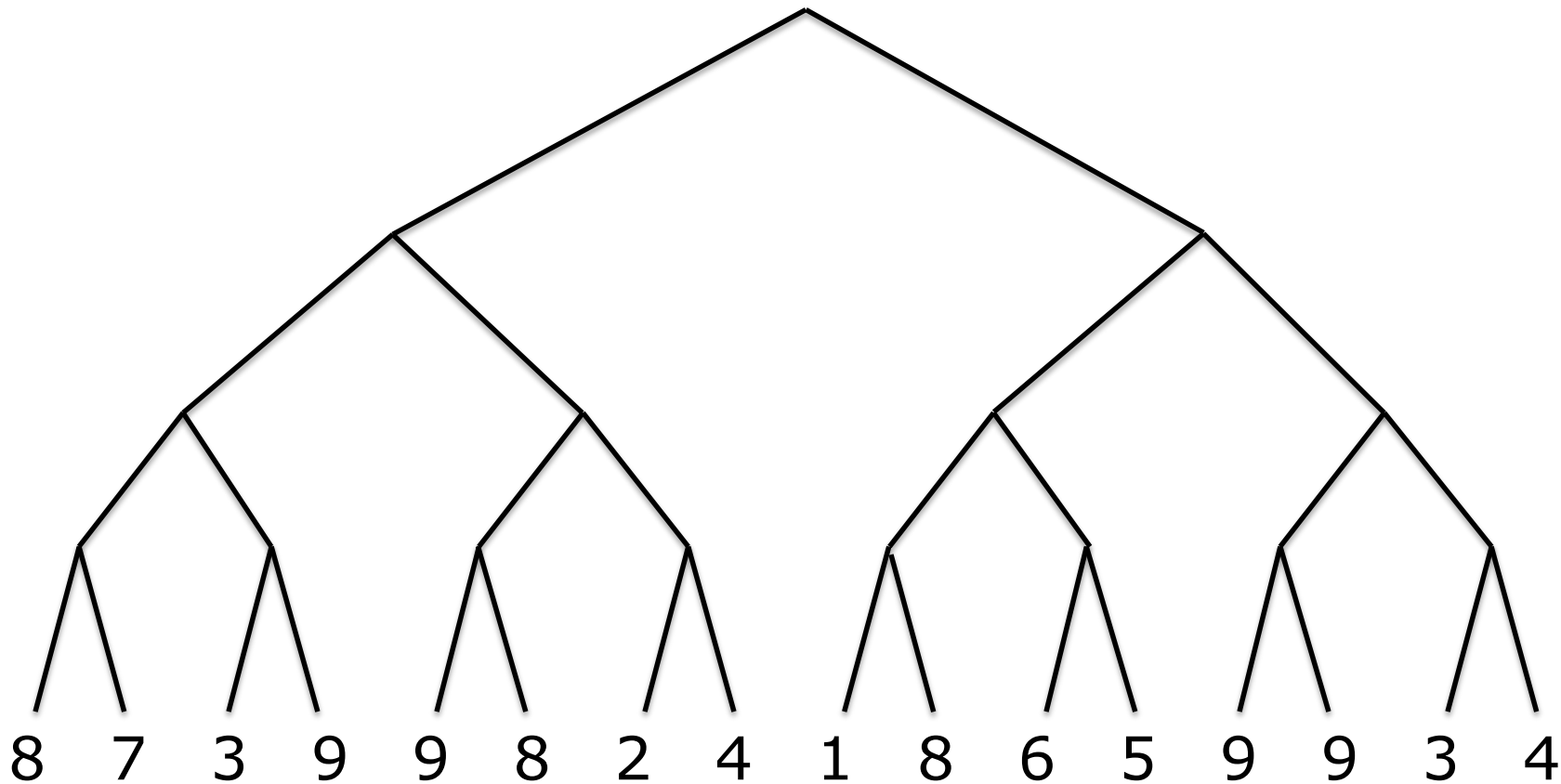
MaxValue(startState, -infinity, +infinity, 0)

Max

Min

Max

Min



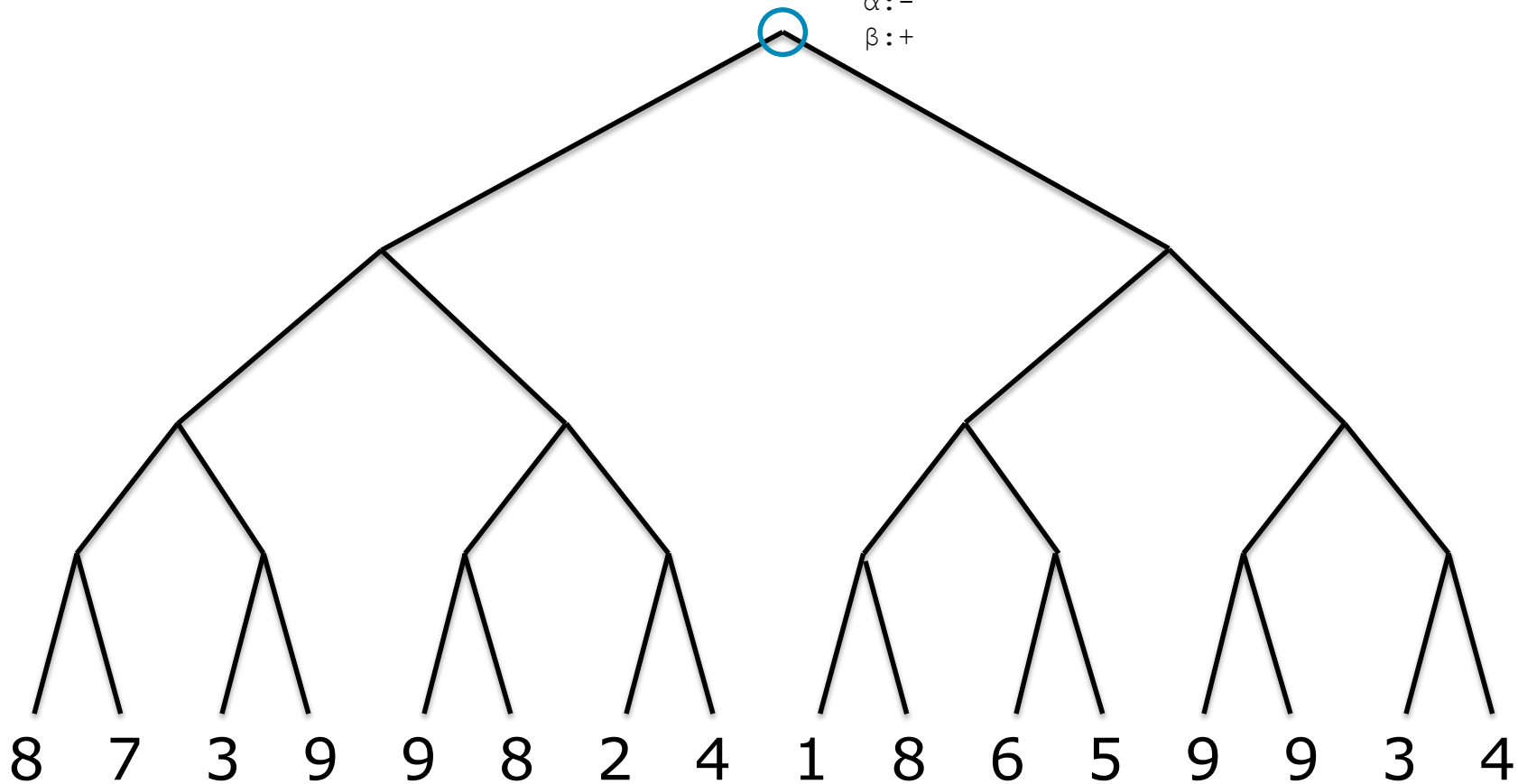
Max

Min

Max

Min

$\forall:-$
 $\alpha:-$
 $\beta:+$

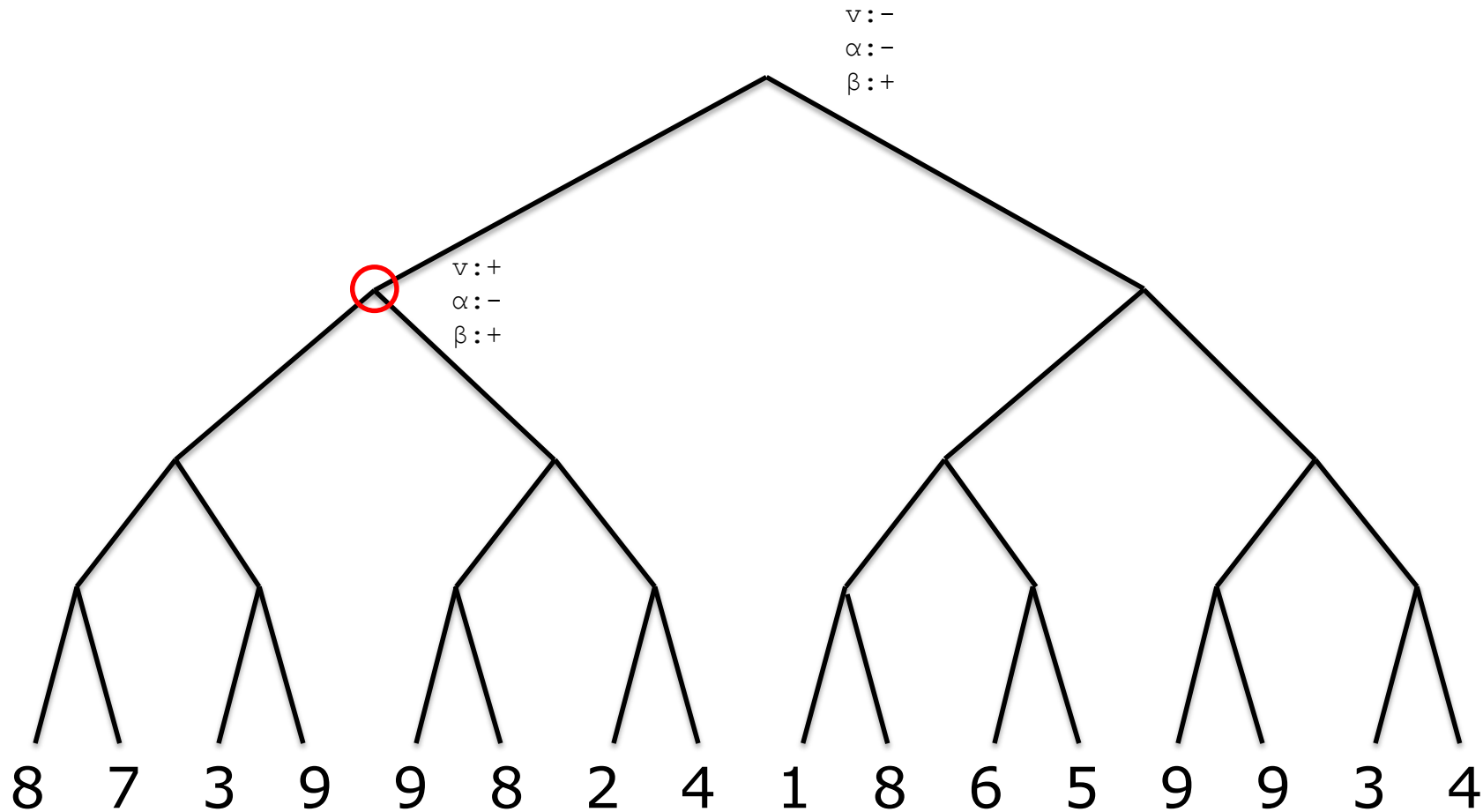


Max

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Max

Min

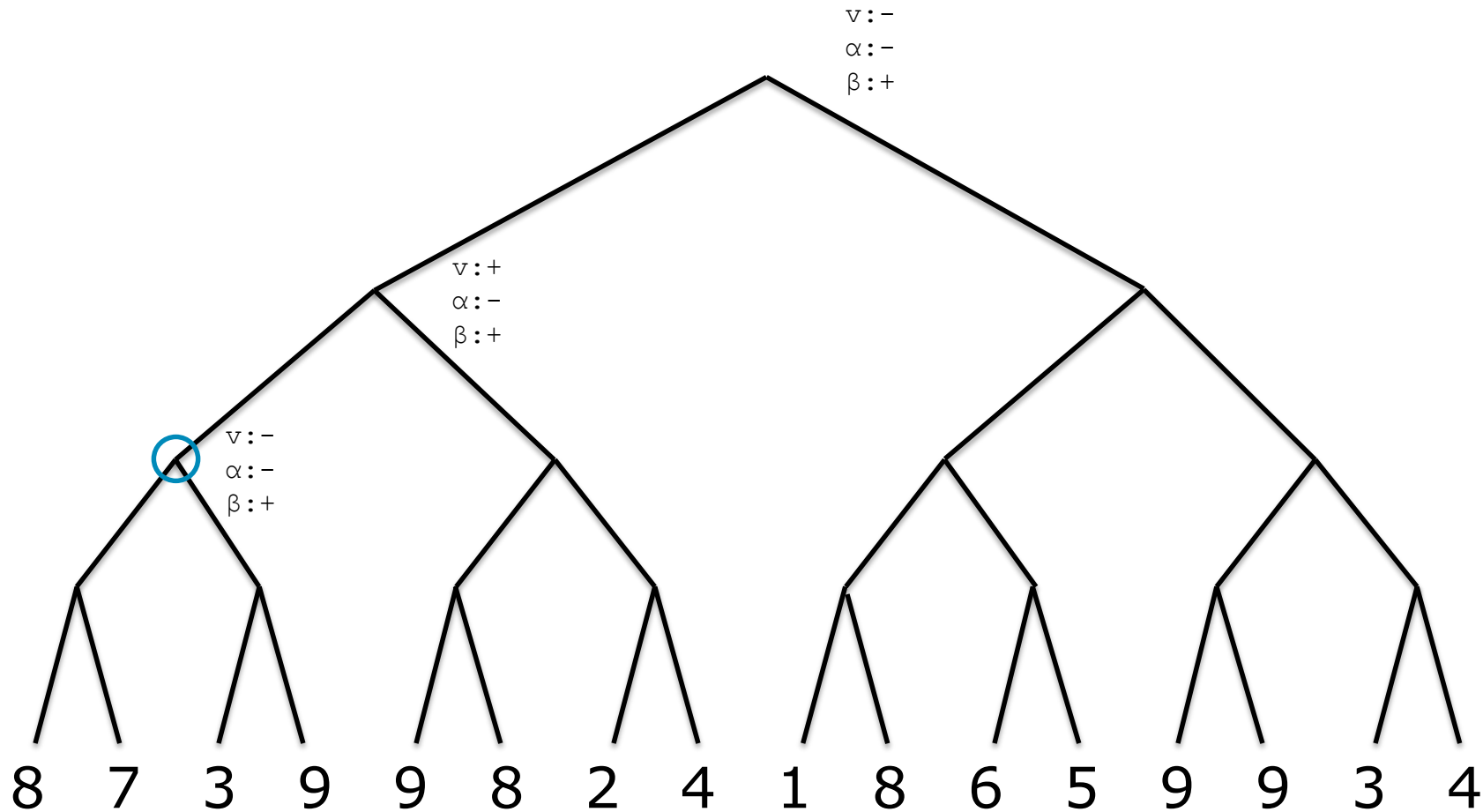


Max

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Max

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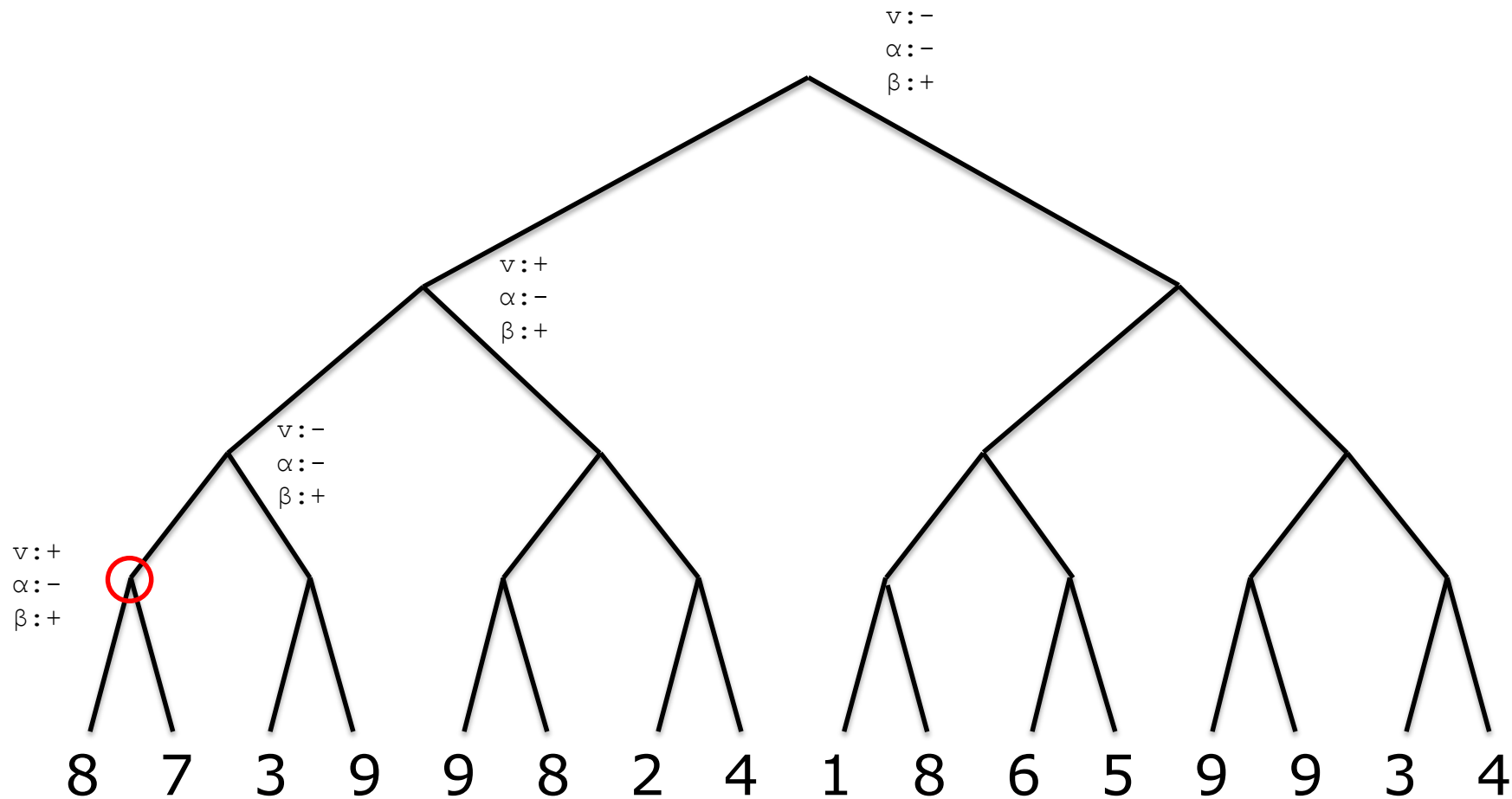


Max

Min

Max

Min



Max

MinValue

if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

Min

Max

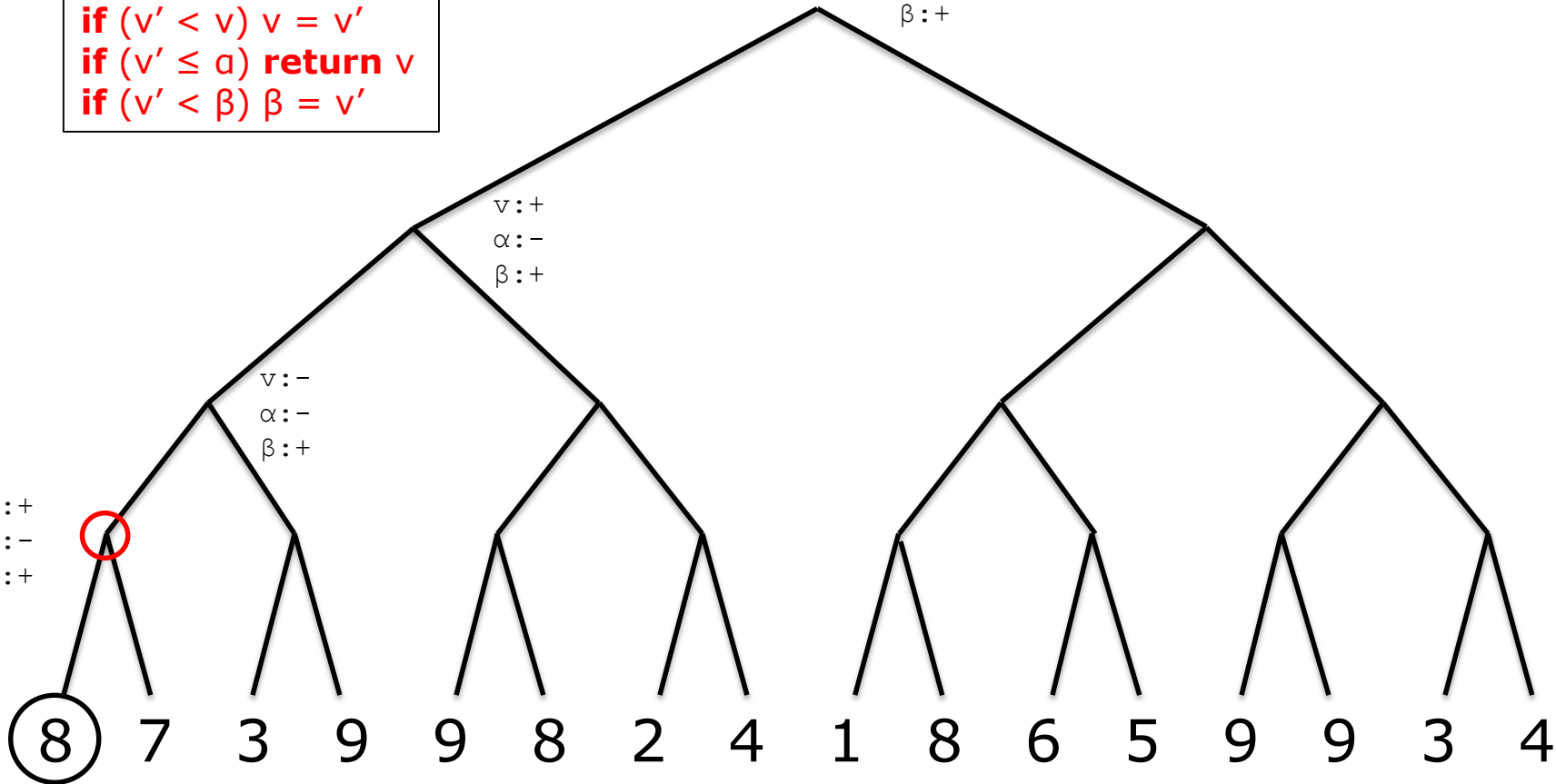
Min

$v:-$
 $\alpha:-$
 $\beta:+$

$v:+$
 $\alpha:-$
 $\beta:+$

$v:-$
 $\alpha:-$
 $\beta:+$

$v:+$
 $\alpha:-$
 $\beta:+$



Max

MinValue

```

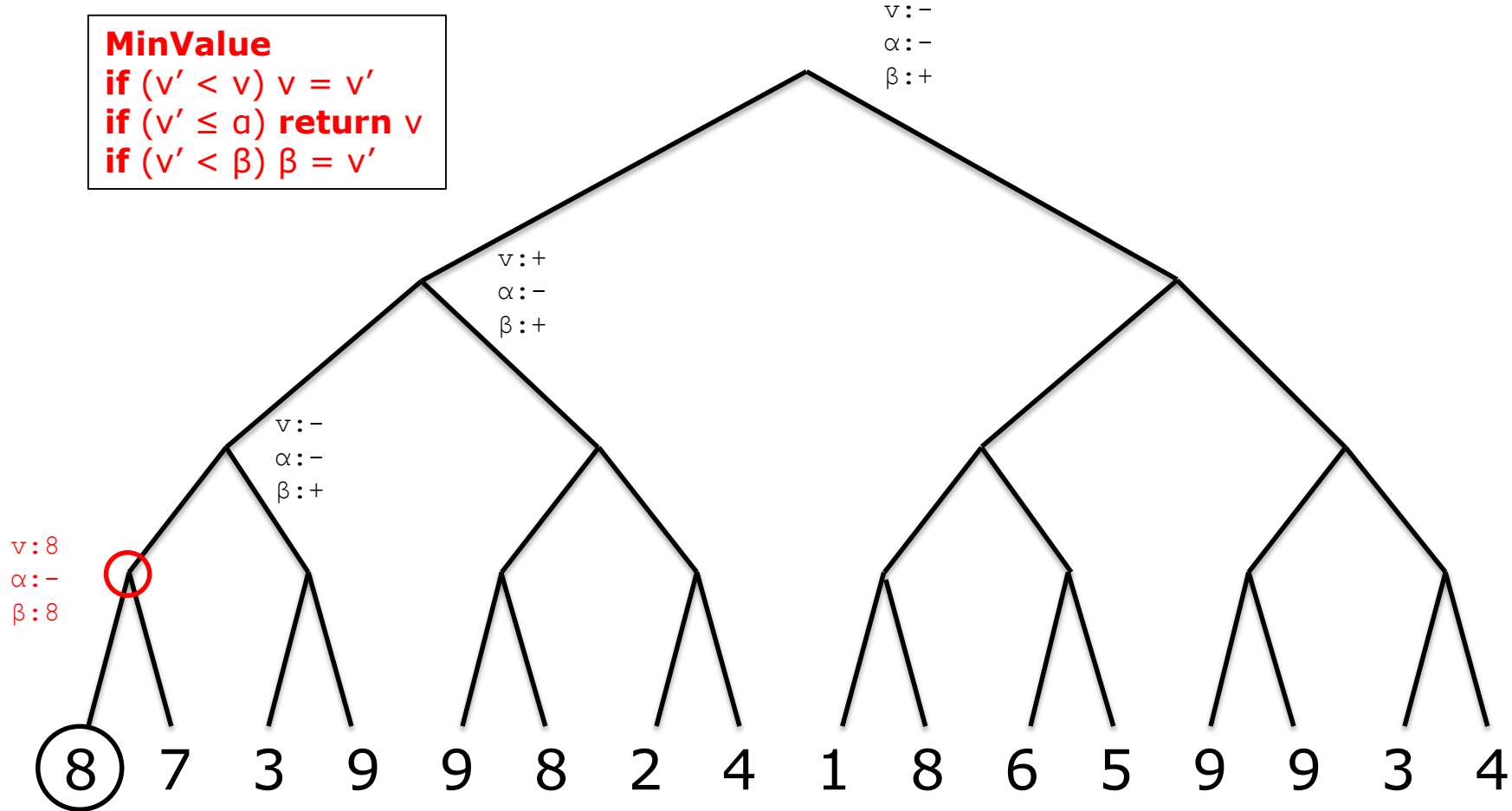
if ( $v' < v$ )  $v = v'$ 
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```

Min

Max

Min



Max

MinValue

if ($v' < v$) $v = v'$

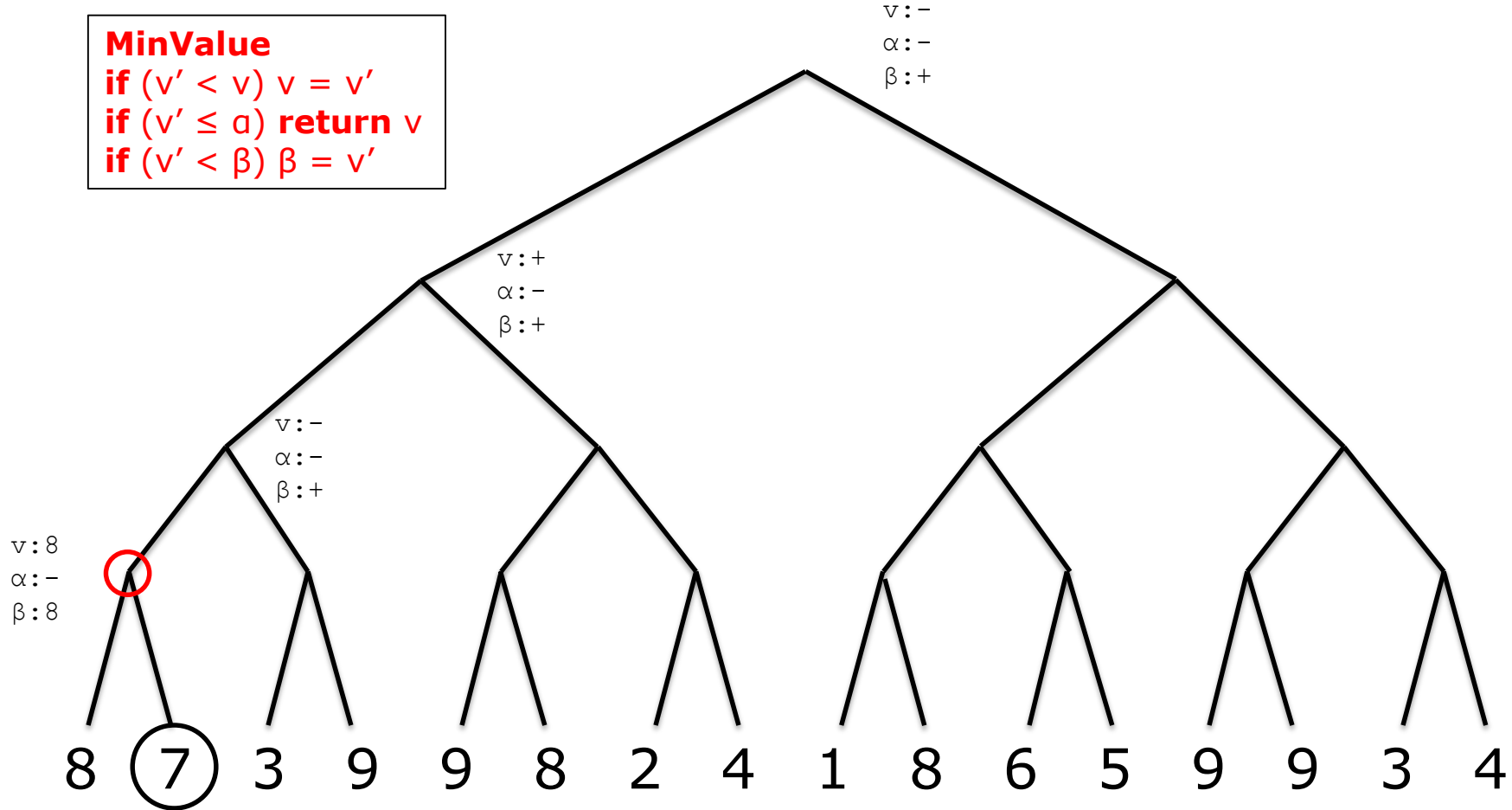
```
if ( $v' \leq a$ ) return  $v$ 
```

if ($v' < \beta$) $\beta = v'$

Min

Max

Min



Max

MinValue

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Min

Max

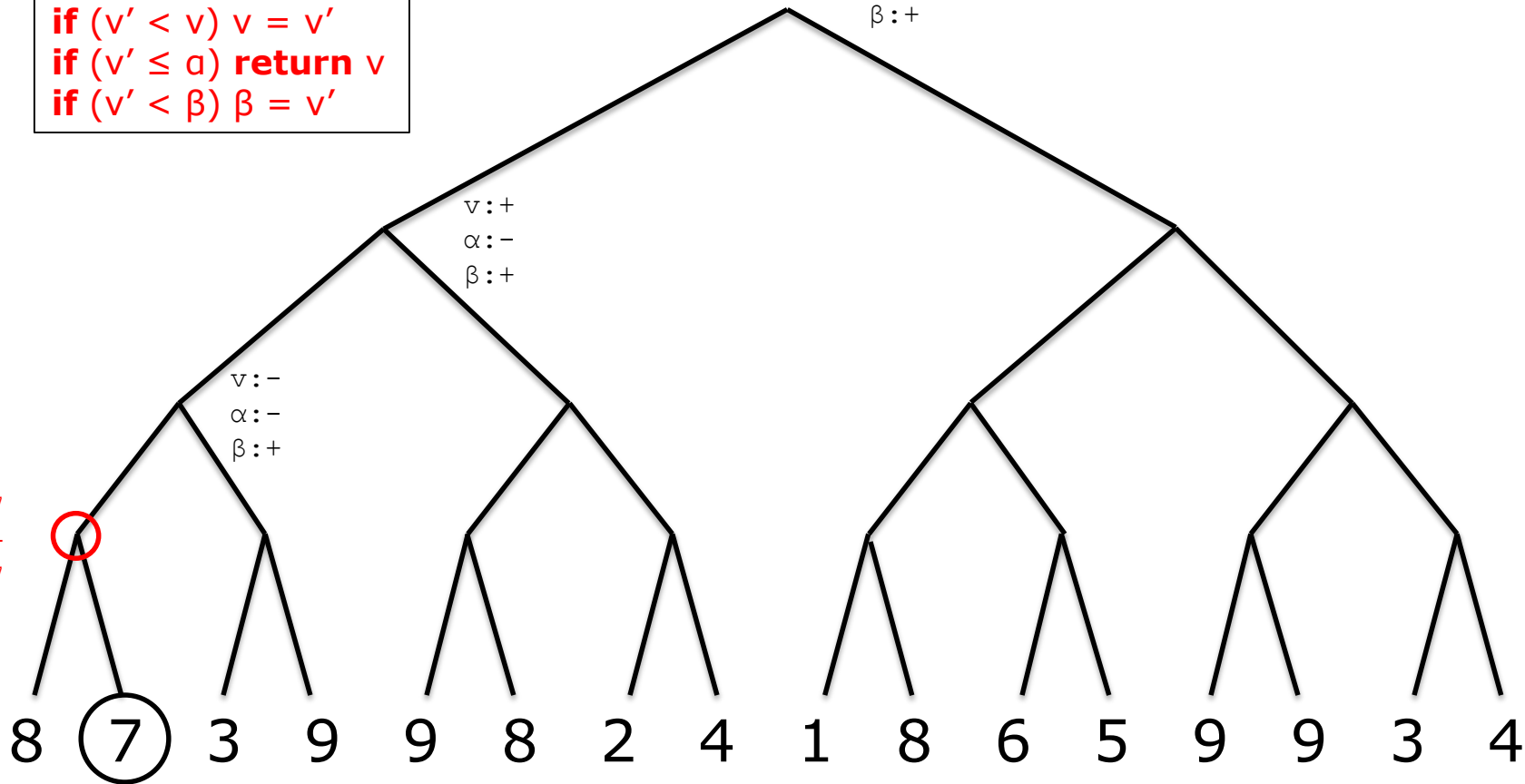
Min

$v:-$
 $\alpha:-$
 $\beta:+$

$v:+$
 $\alpha:-$
 $\beta:+$

$v:-$
 $\alpha:-$
 $\beta:+$

$v:7$
 $\alpha:-$
 $\beta:7$



Max

MinValue

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Min

Max

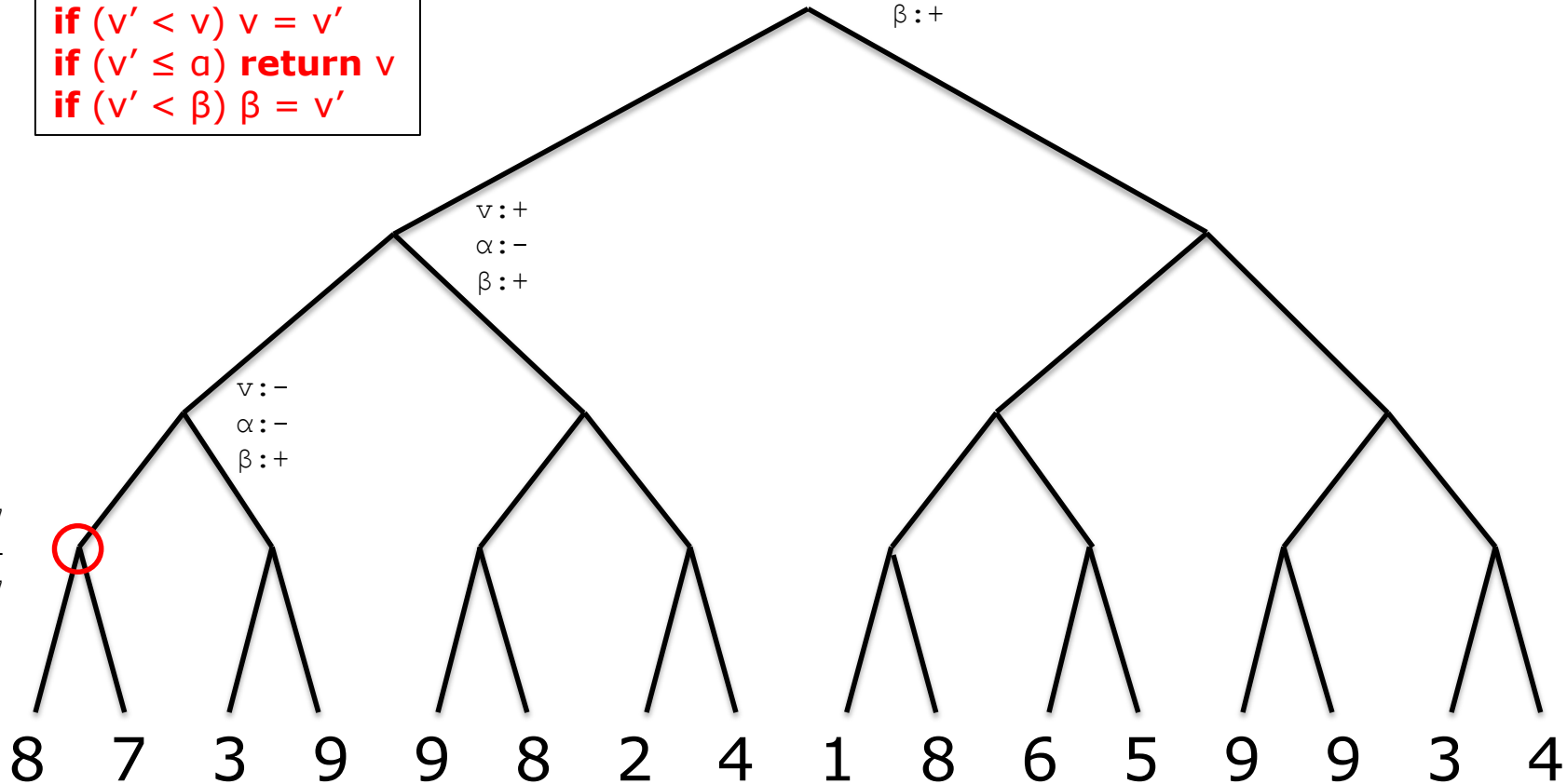
Min

$v:-$
 $\alpha:-$
 $\beta:+$

$v:+$
 $\alpha:-$
 $\beta:+$

$v:-$
 $\alpha:-$
 $\beta:+$

$v:7$
 $\alpha:-$
 $\beta:7$



Max

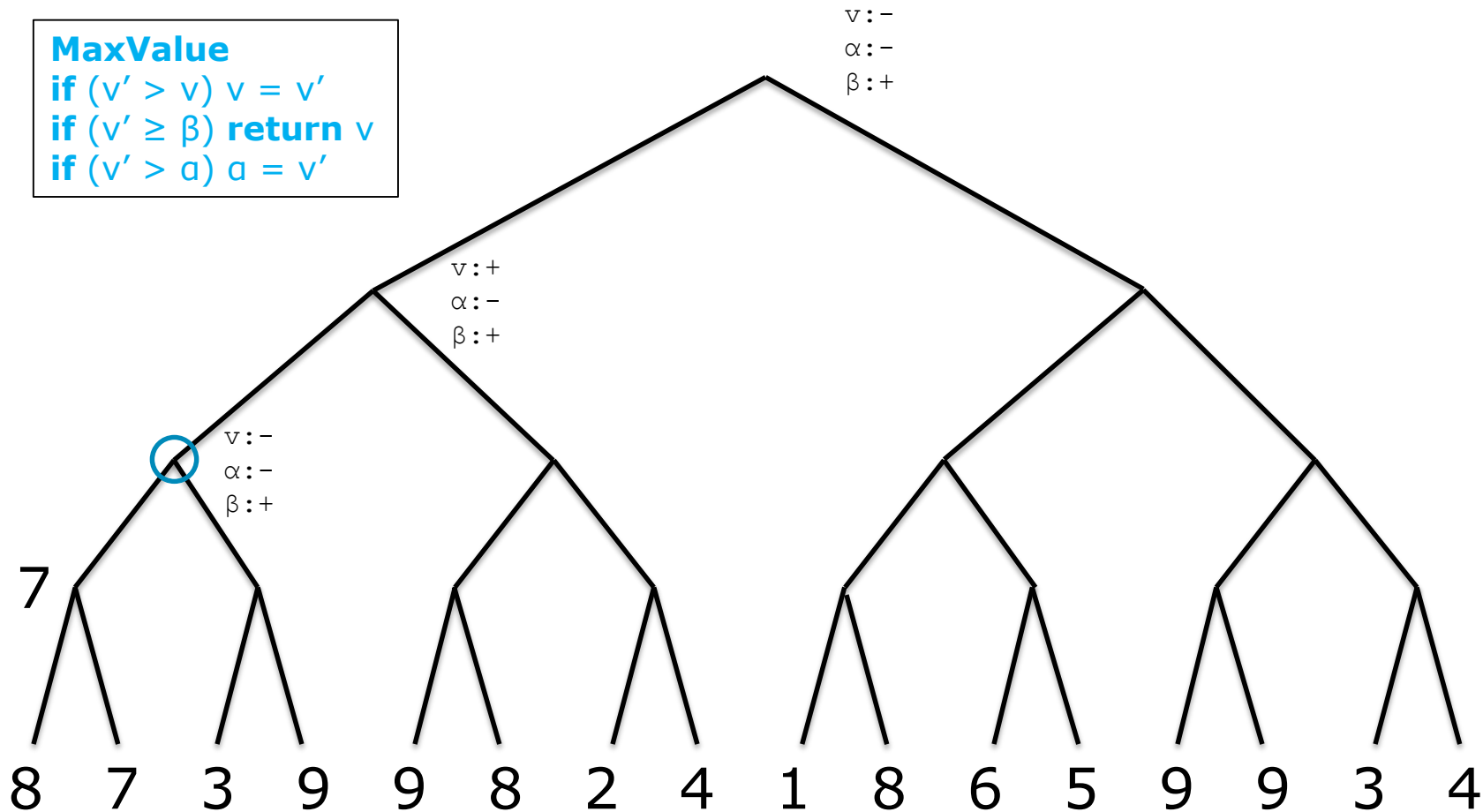
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



Max

MaxValue

```

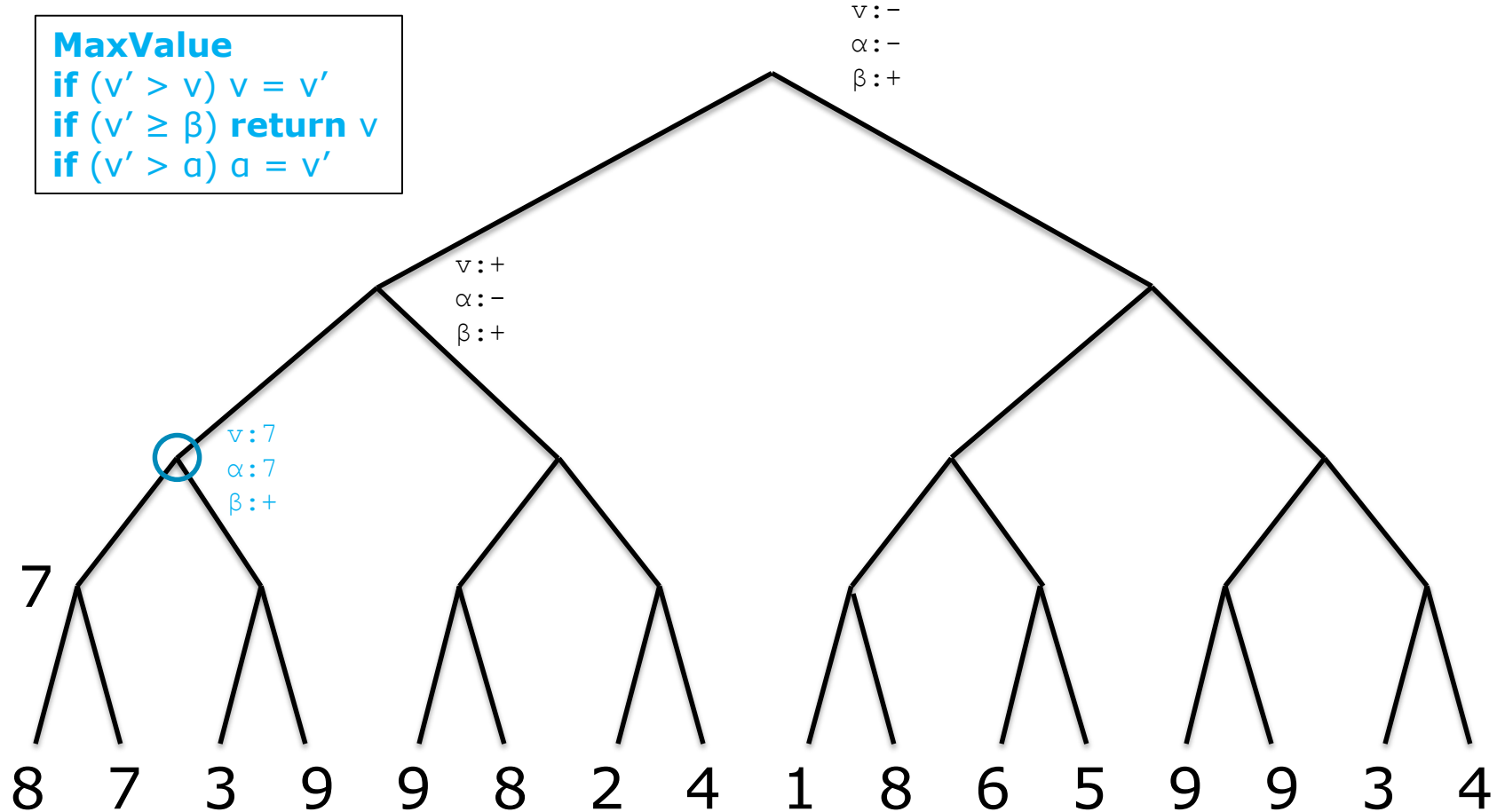
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```

Min

Max

Min



Max

MaxValue

```

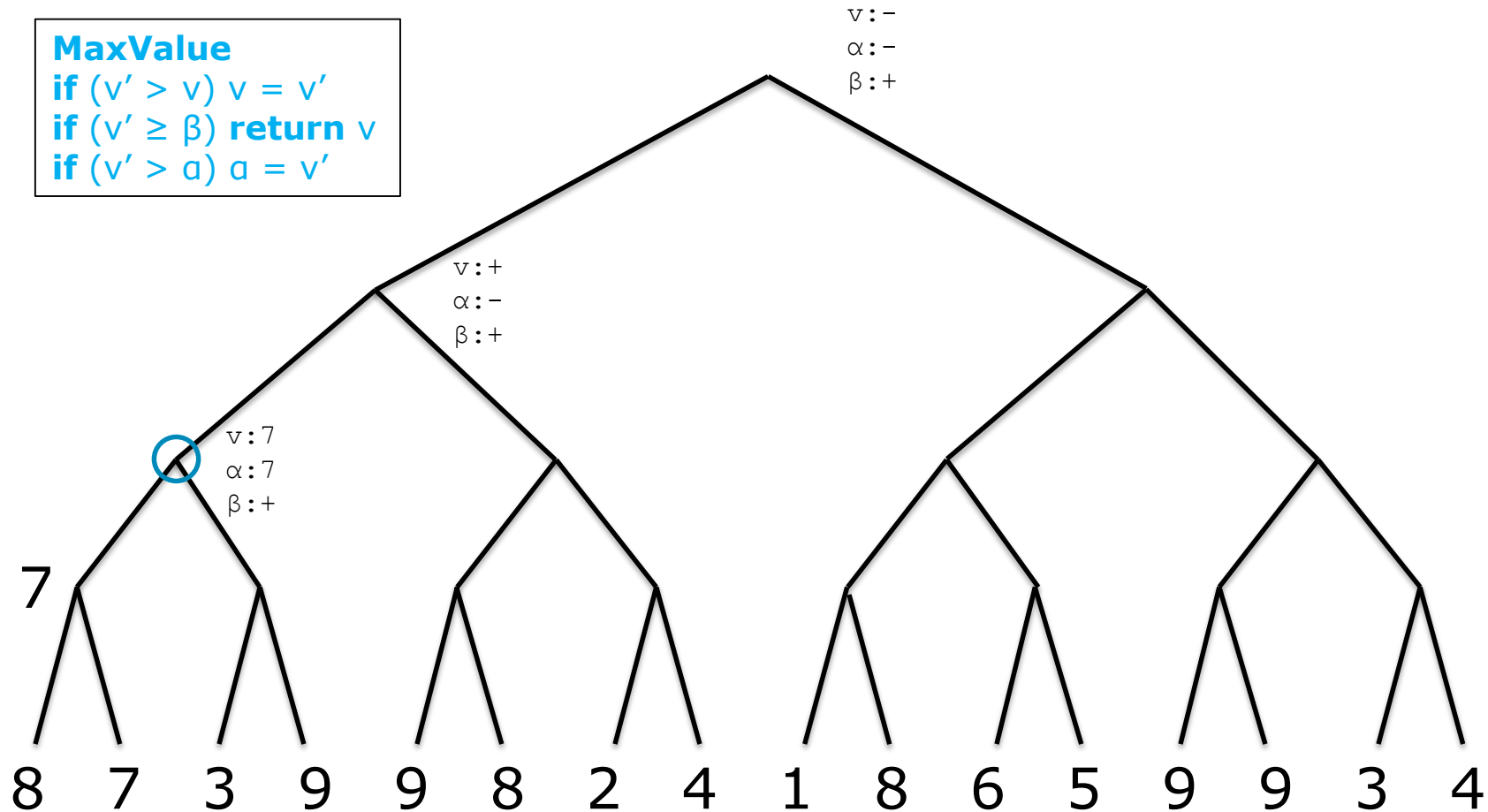
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```

Min

Max

Min

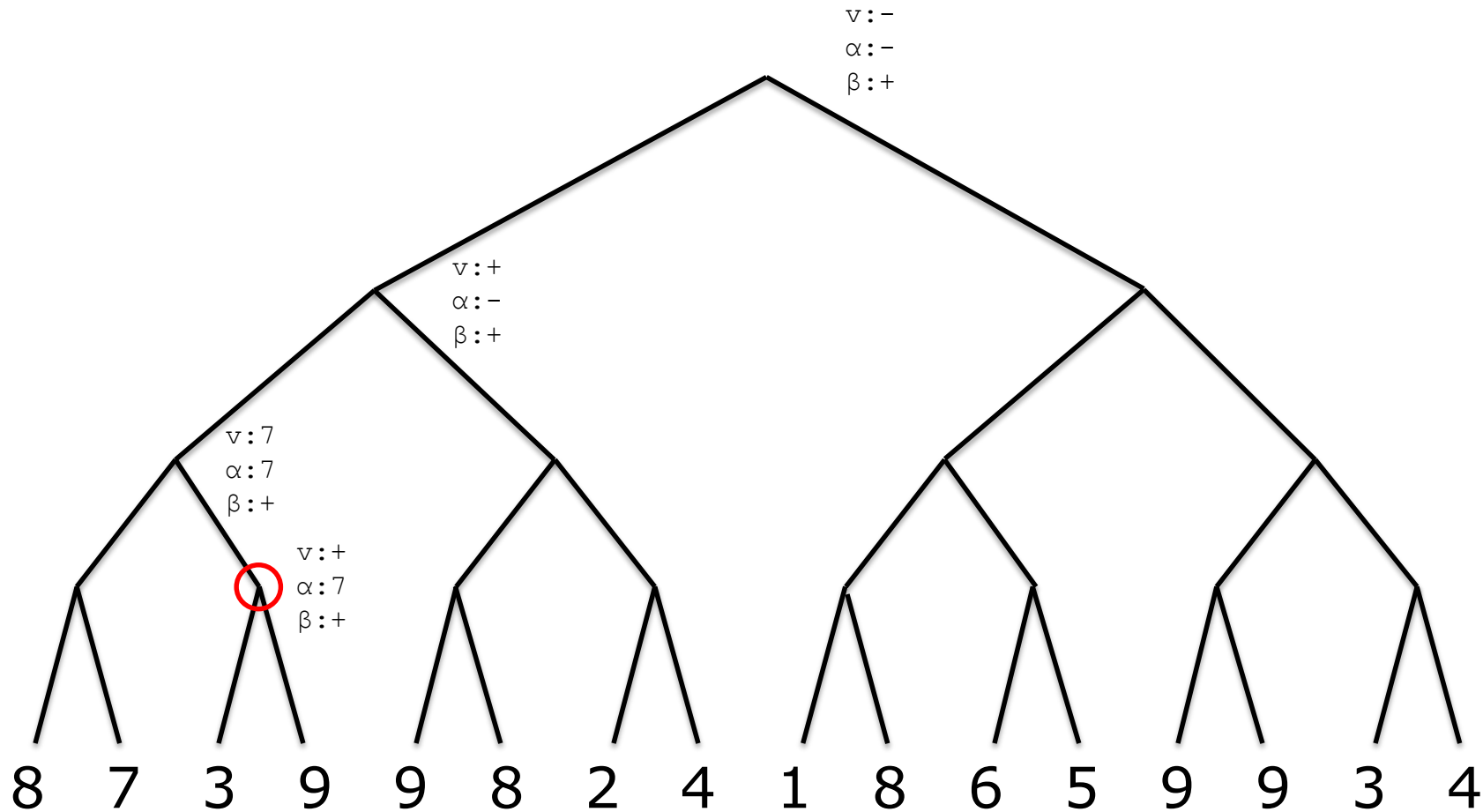


Max

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Max

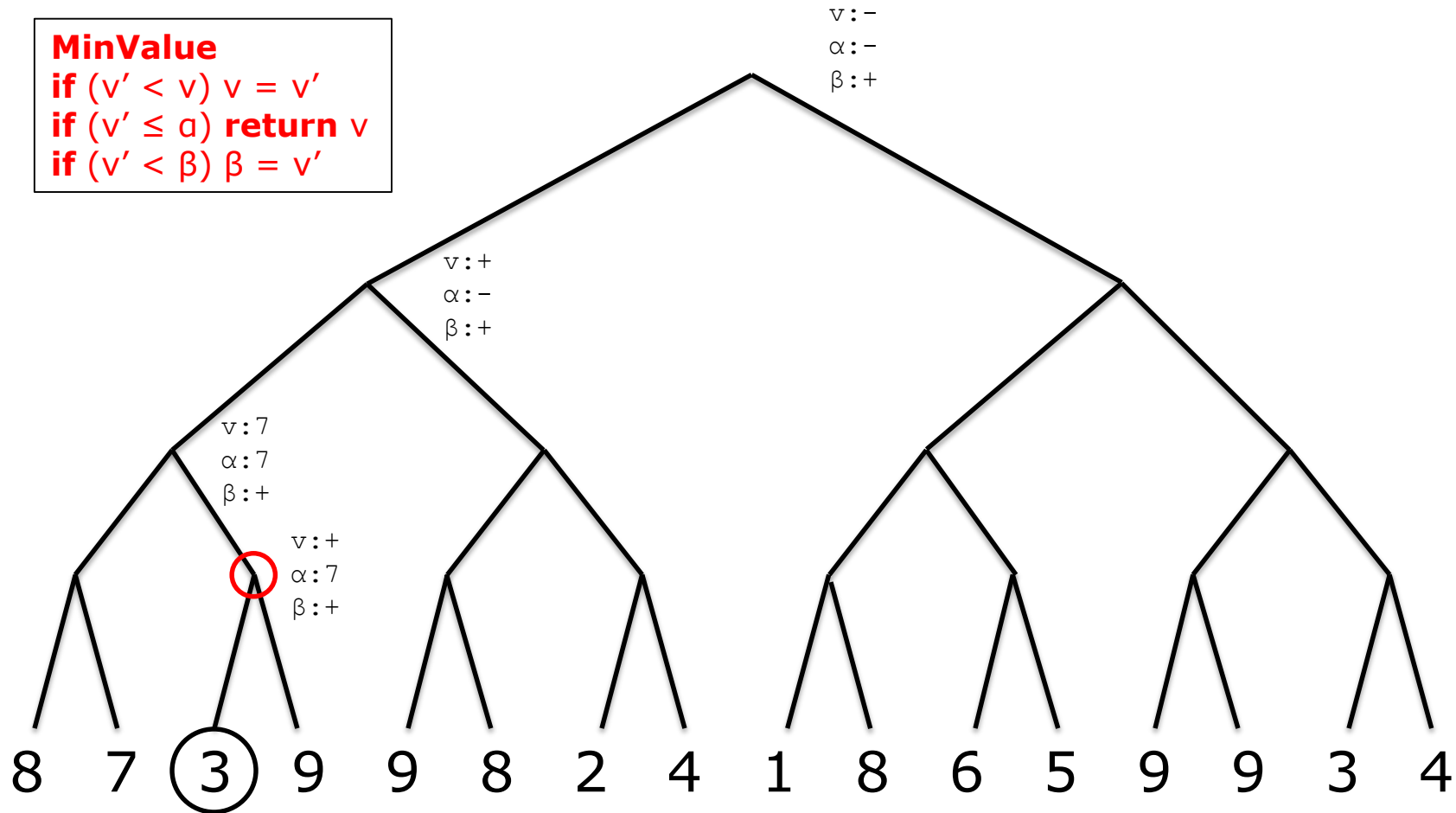
MinValue

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if ($v' < \beta$) $\beta = v'$

Min

Max

Min



Max

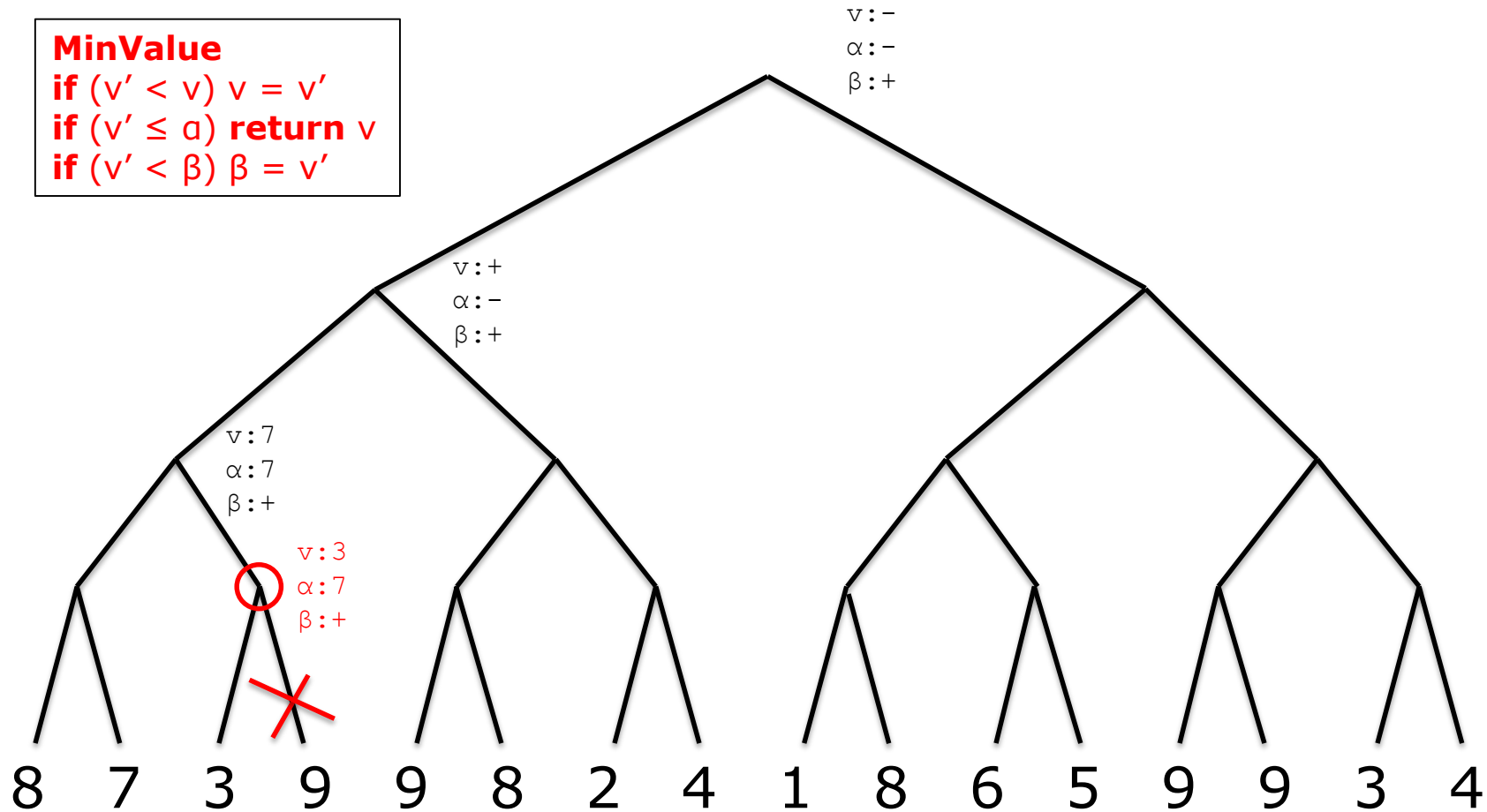
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Max

Min



Max

MaxValue

```

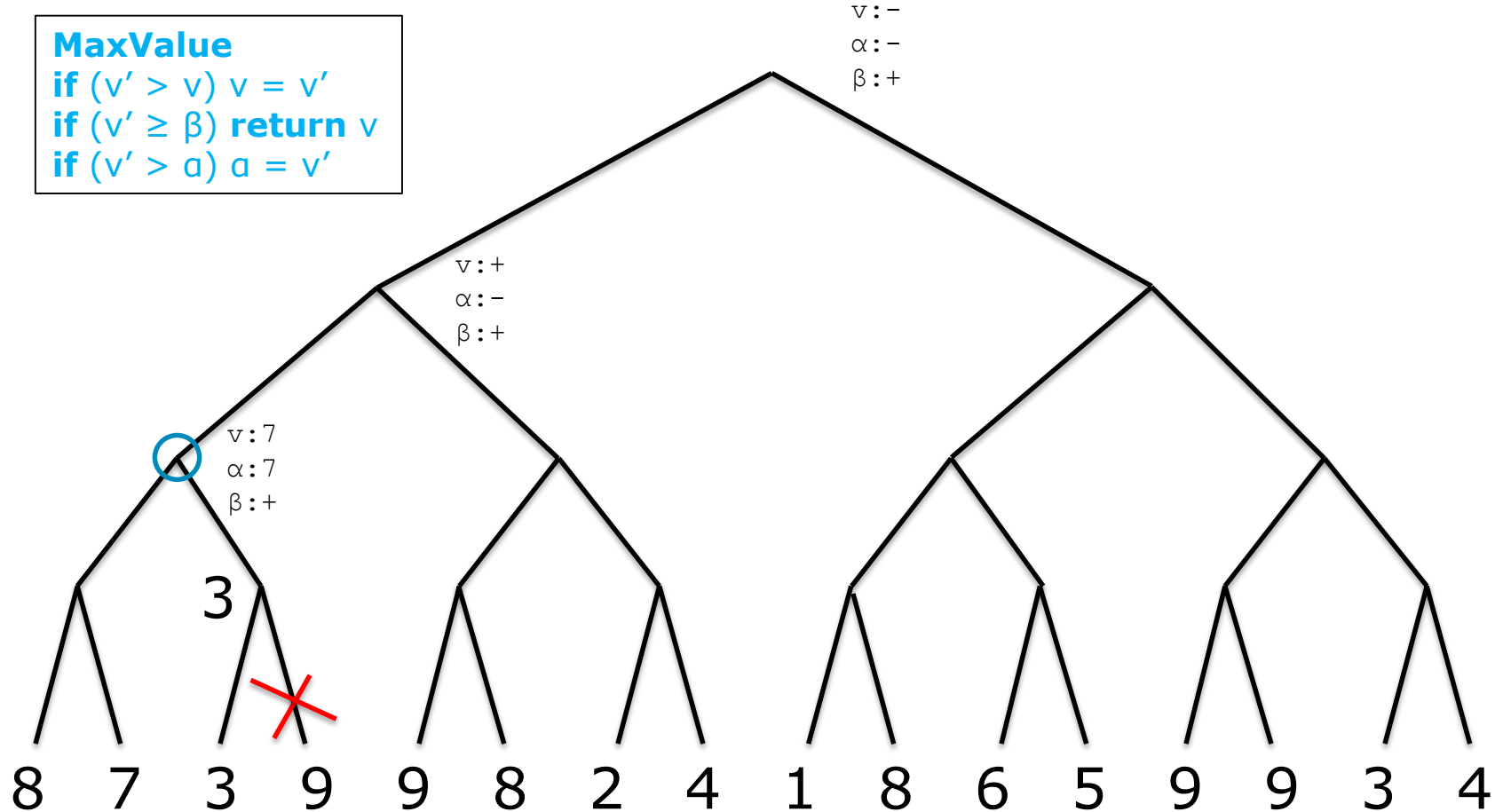
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```

Min

Max

Min



Max

MaxValue

```

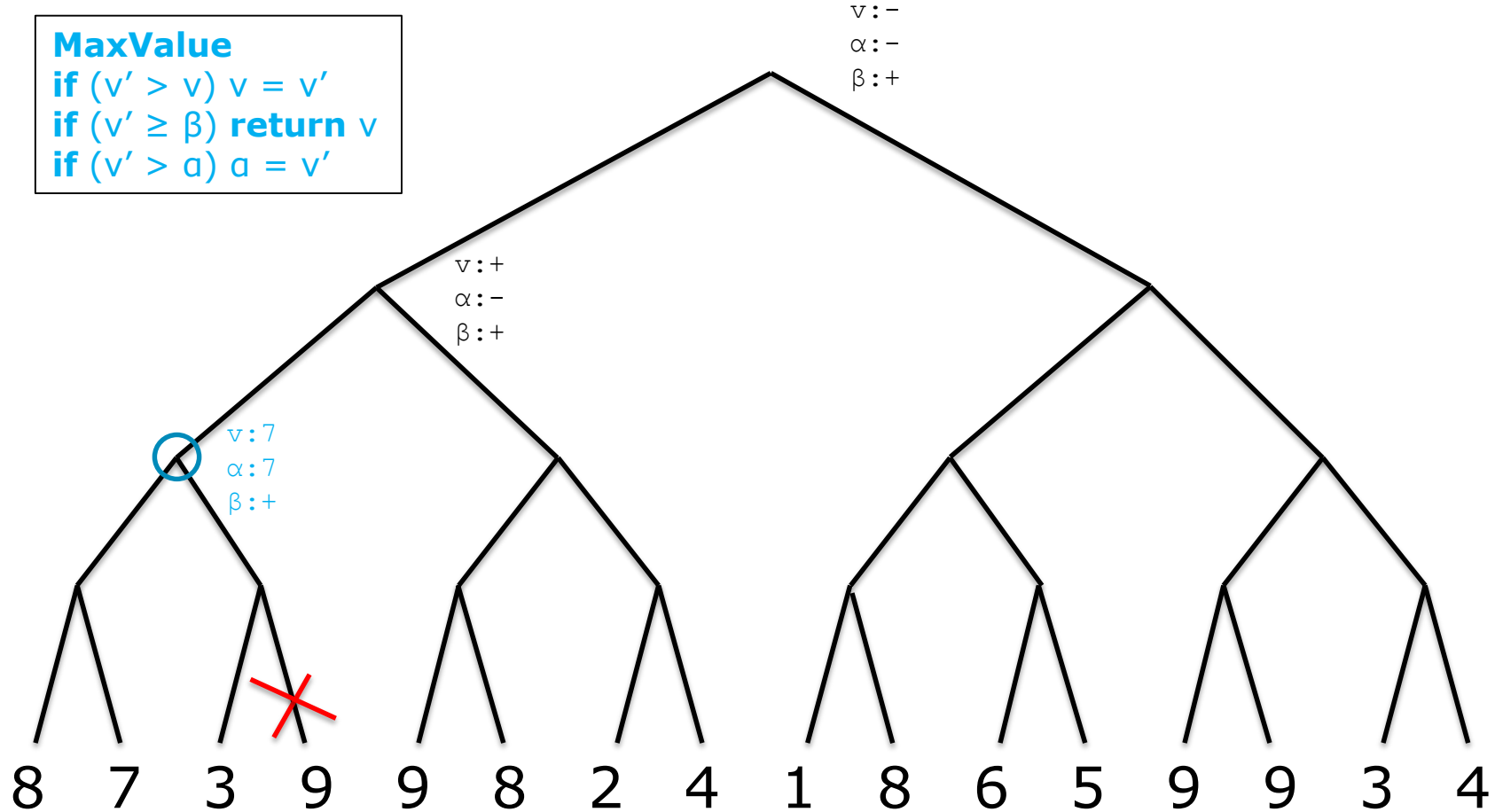
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```

Min

Max

Min



Max

MinValue

if ($v' < v$) $v = v'$

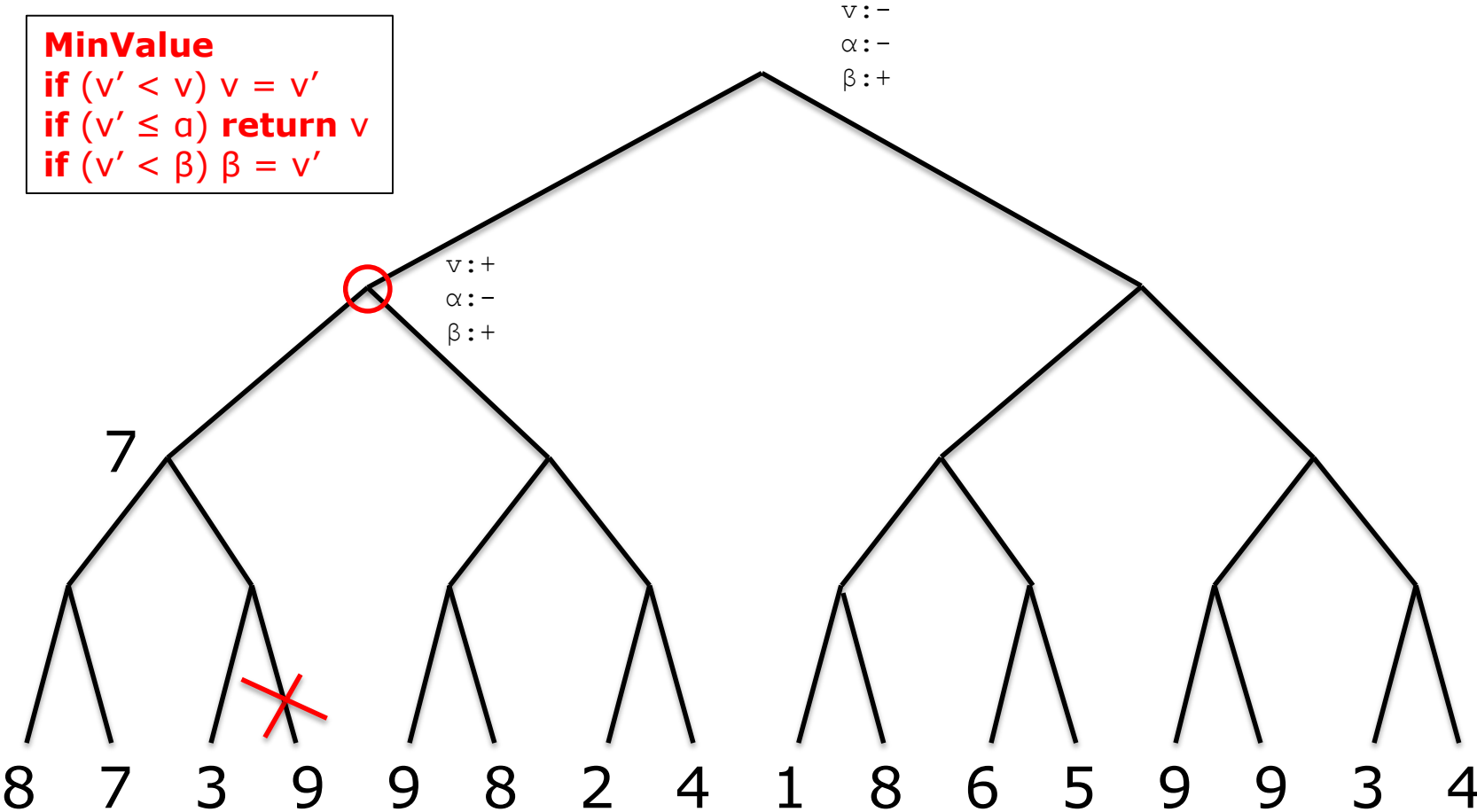
if ($v' \leq a$) return v

if ($v' < \beta$) $\beta = v'$

Min

Max

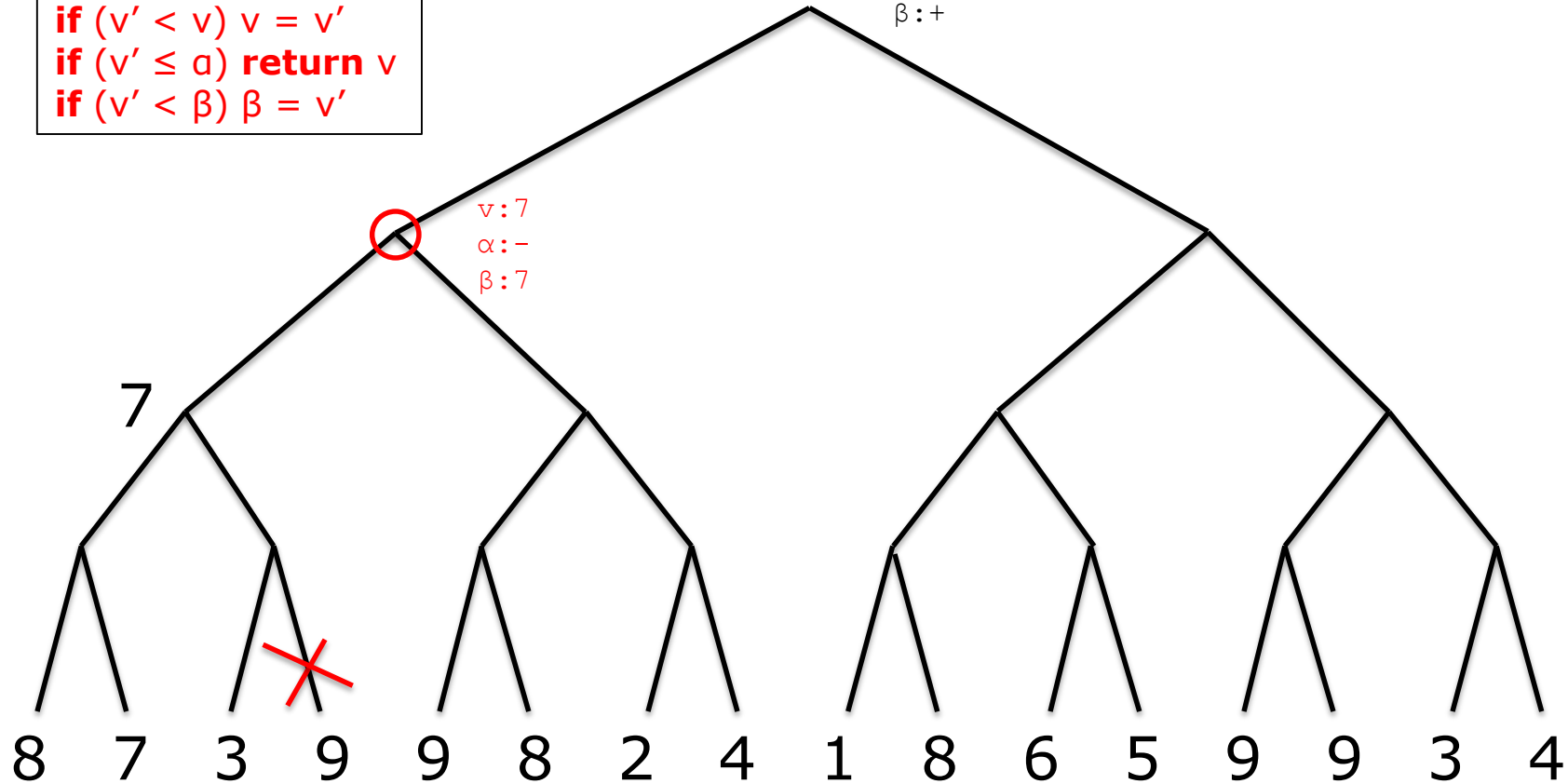
Mir



```

if ( $v' < v$ )  $v = v'$ 
if ( $v' \leq \alpha$ ) return  $v$ 
if ( $v' < \beta$ )  $\beta = v'$ 

```

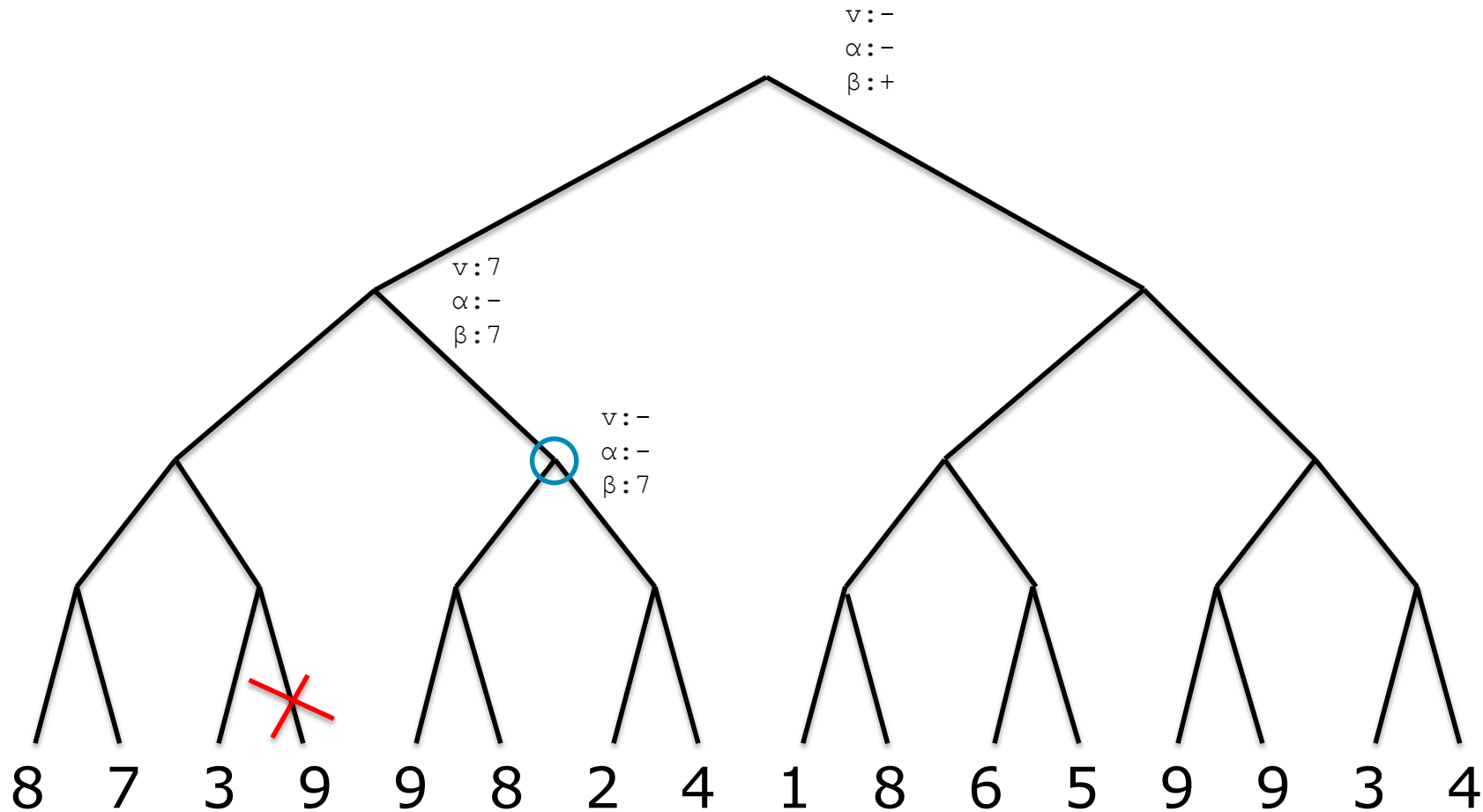
$$\begin{array}{l} \nu : - \\ \alpha : - \\ \beta : + \end{array}$$


Max

Min

Max

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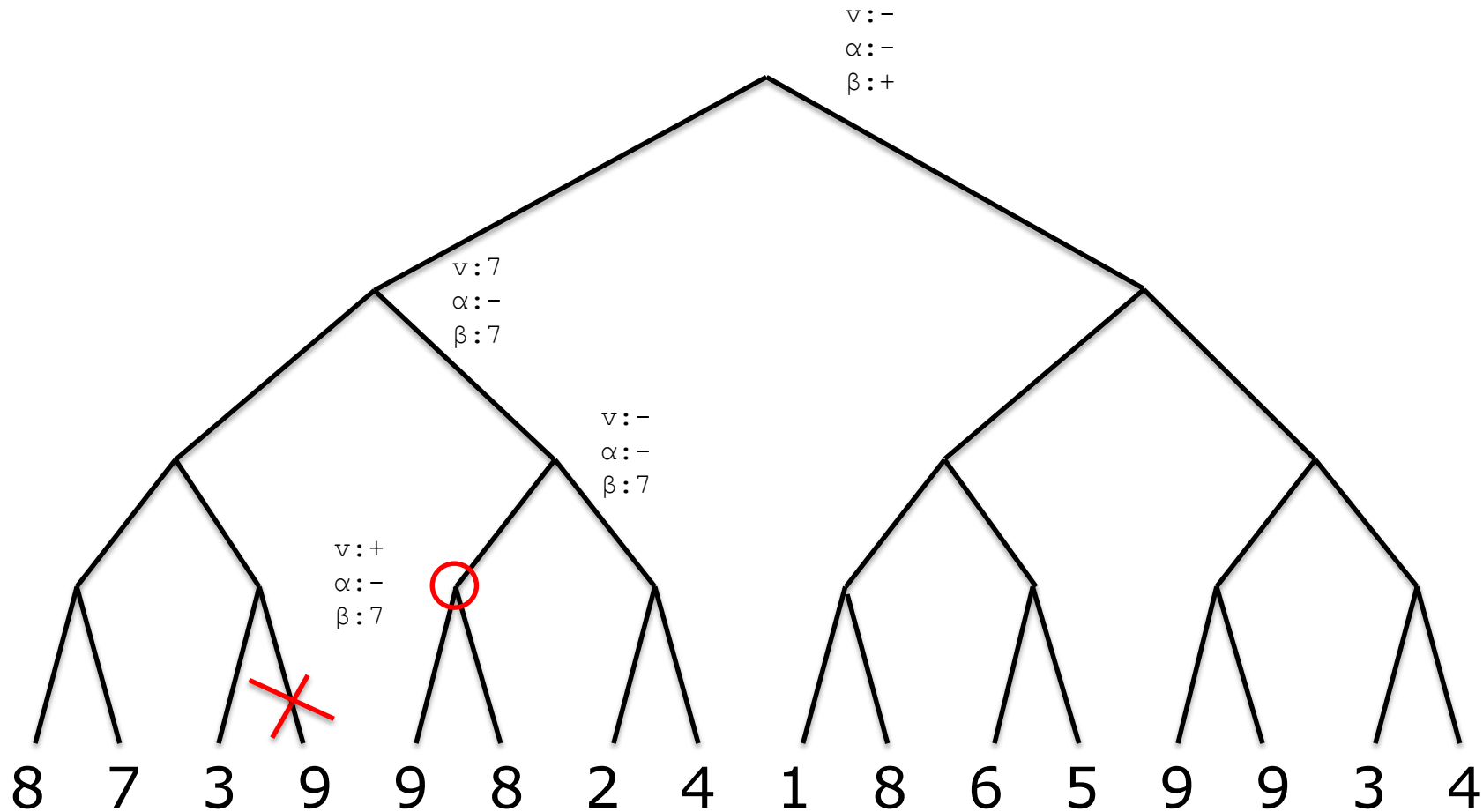


Max

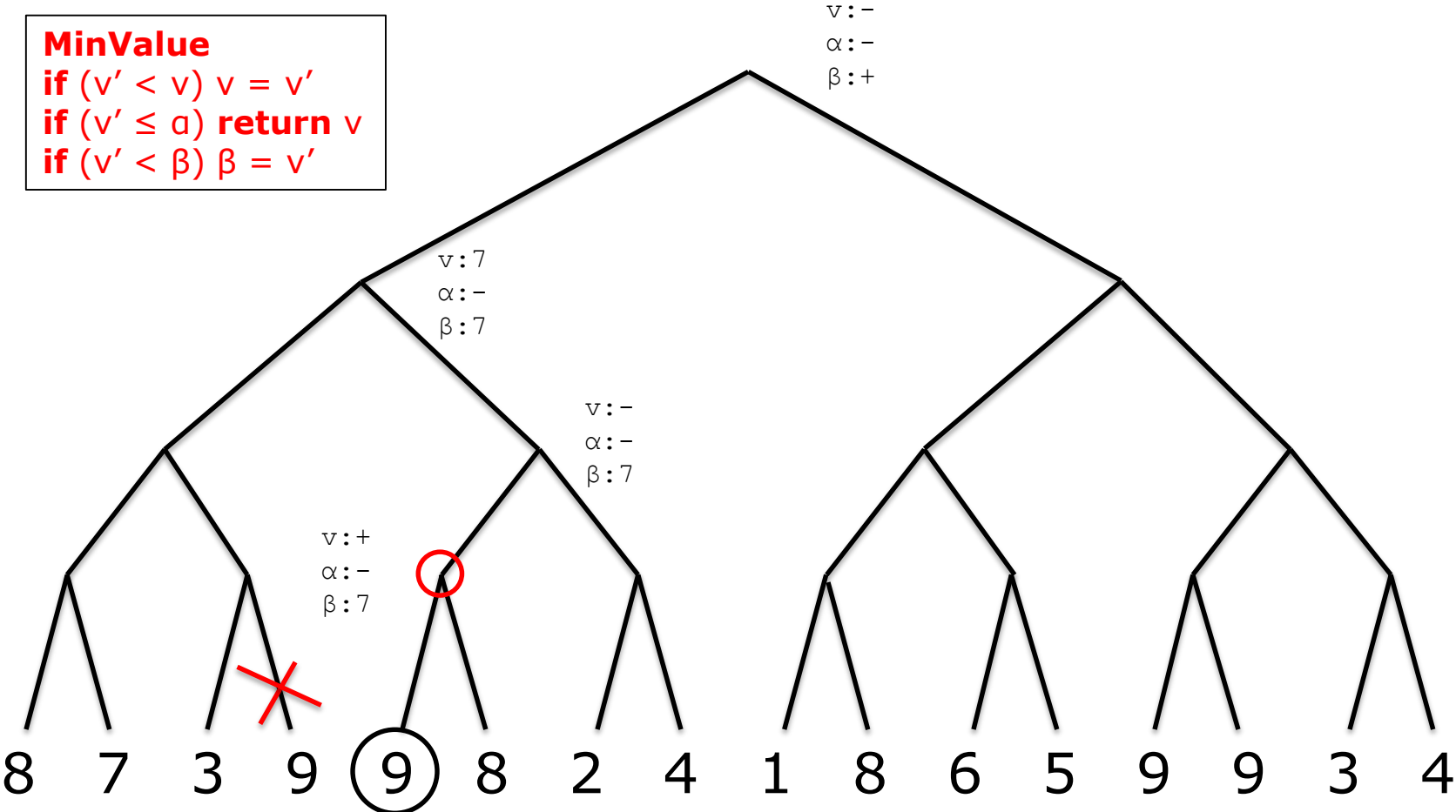
Min

Max

Min



MinValue
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if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$



Max

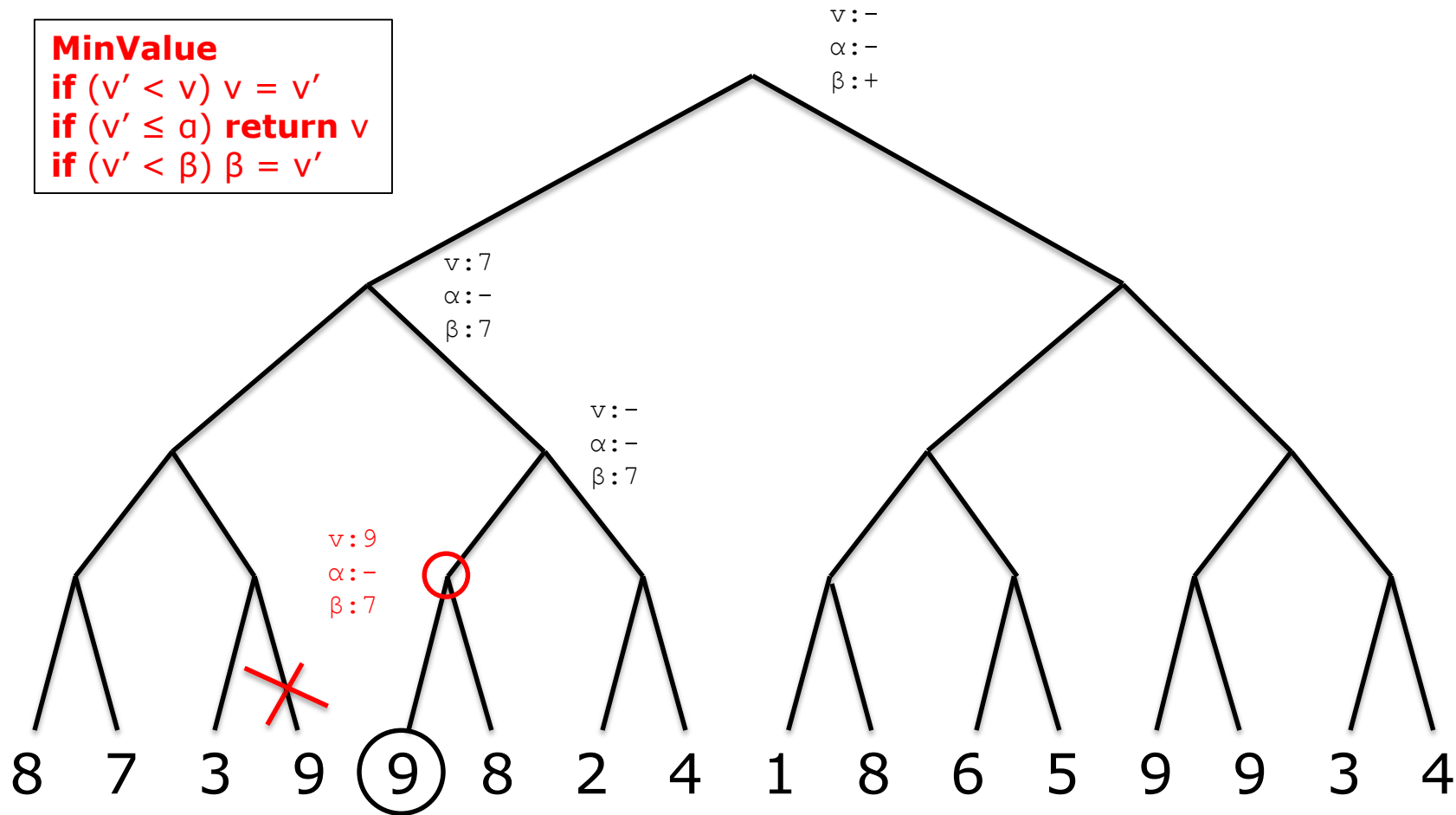
MinValue

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if ($v' < \beta$) $\beta = v'$

Min

Max

Min



```

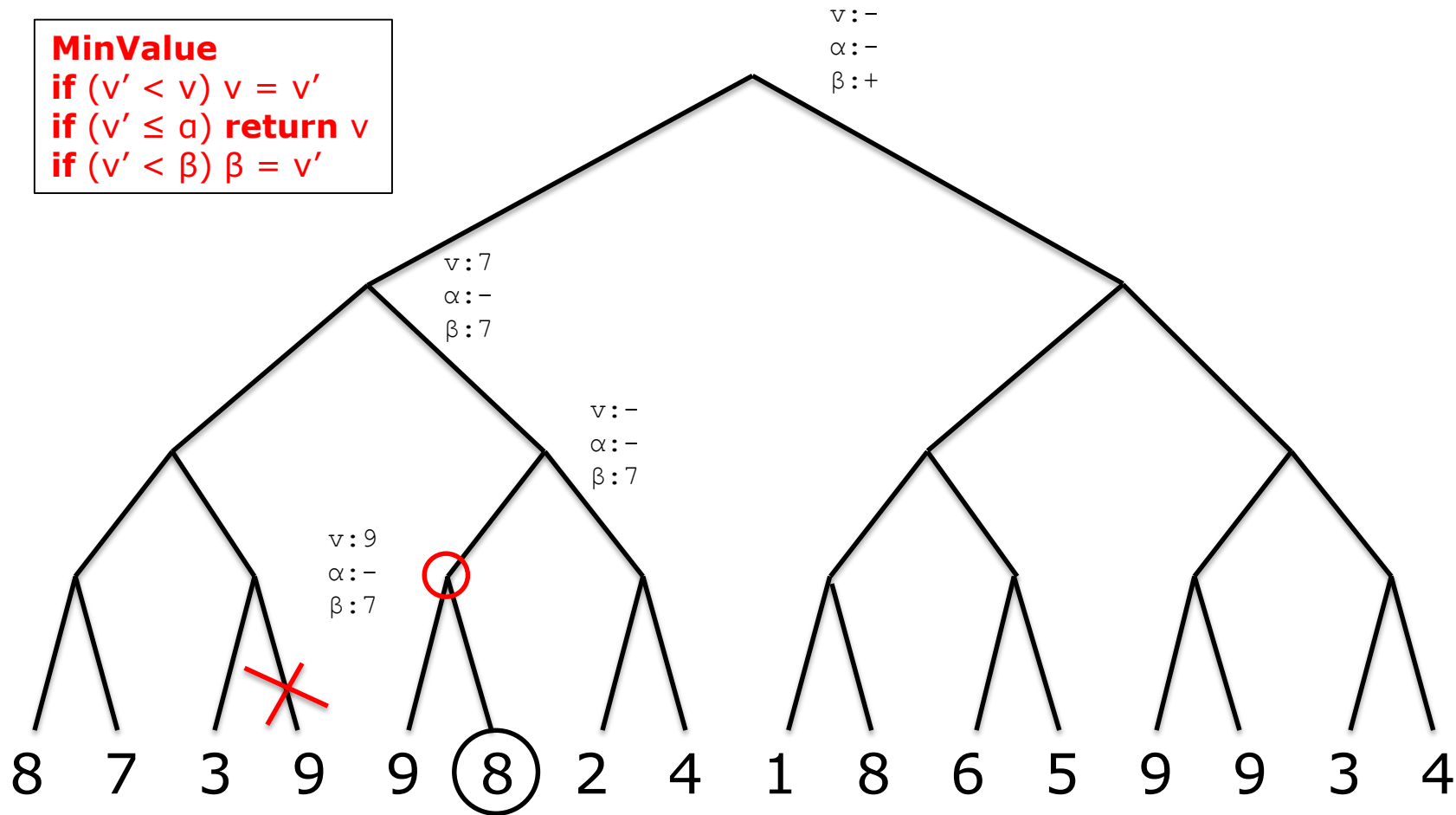
if ( $v' < v$ )  $v = v'$ 
if ( $v' \leq \alpha$ ) return  $v$ 
if ( $v' < \beta$ )  $\beta = v'$ 

```

Min

Max

Min



Max

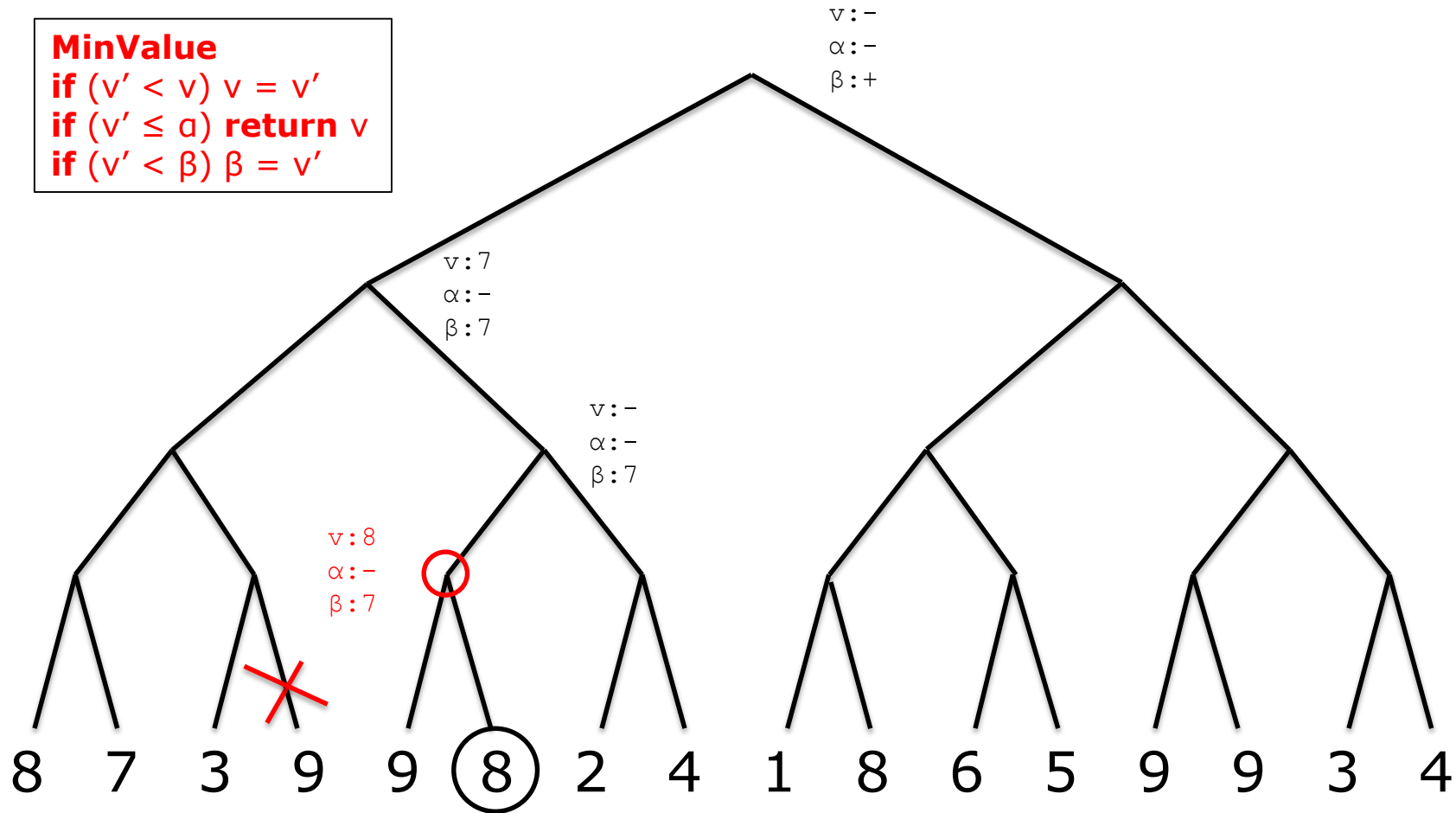
MinValue

if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

Min

Max

Min



Max

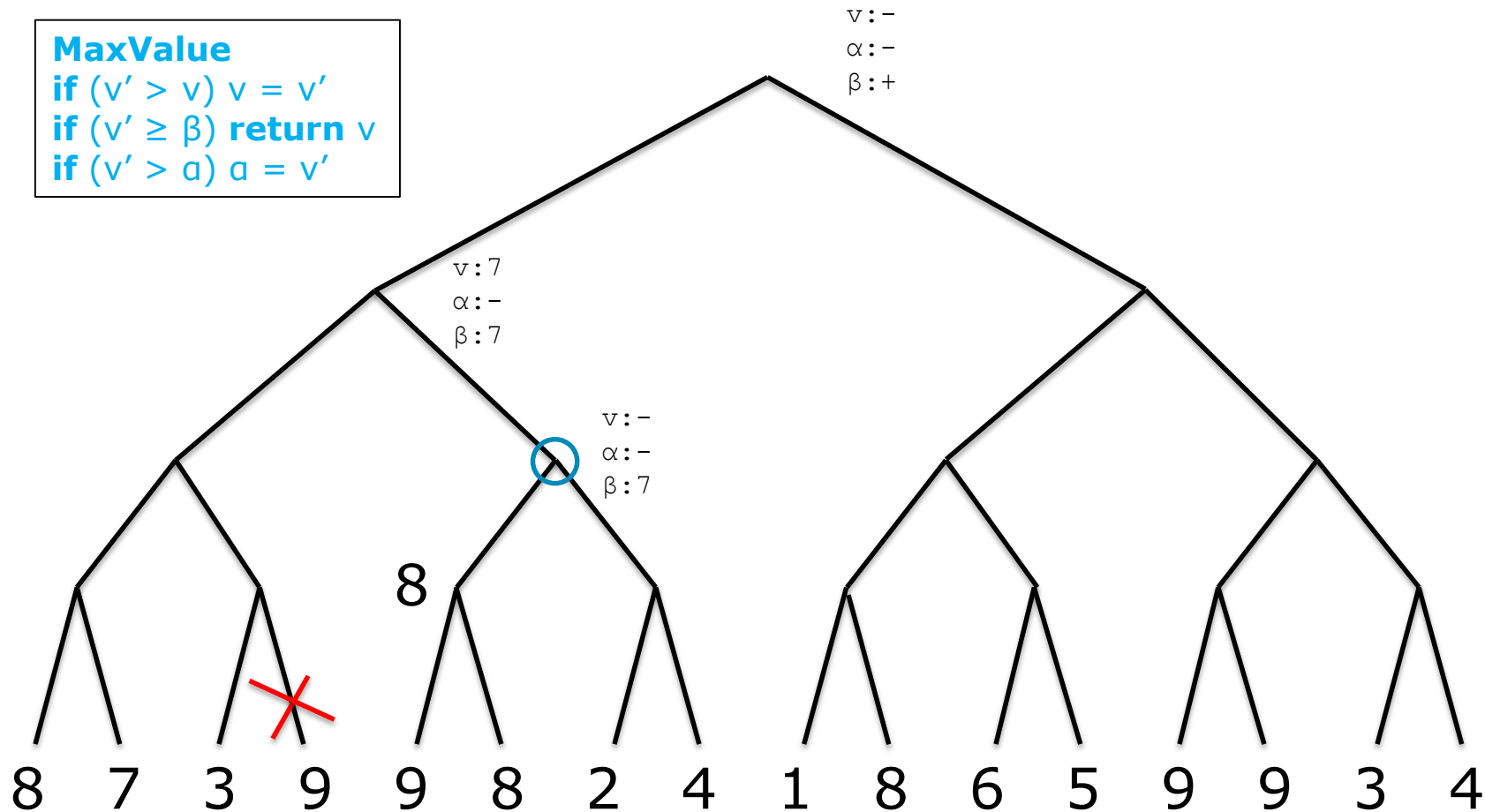
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



Max

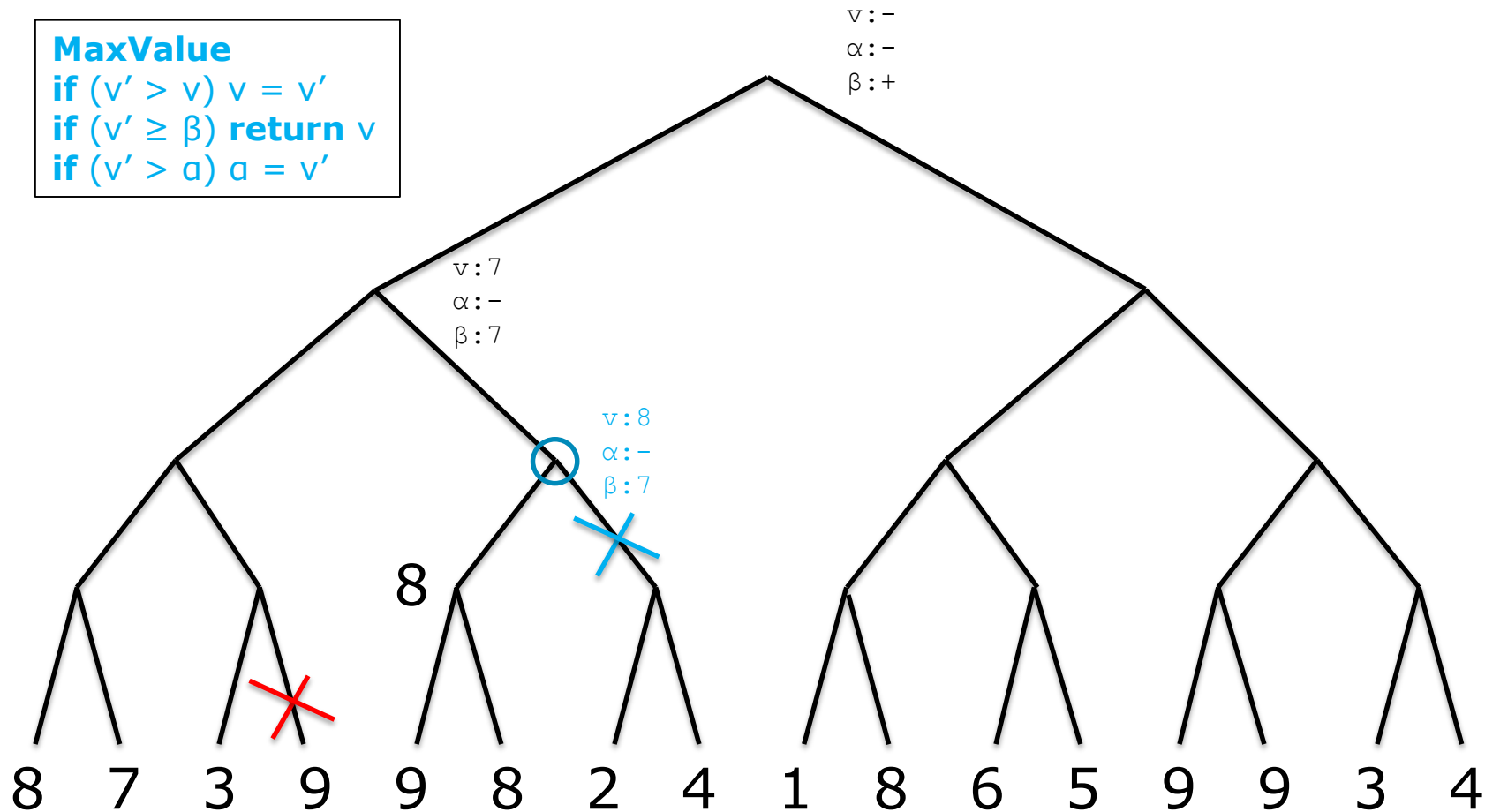
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



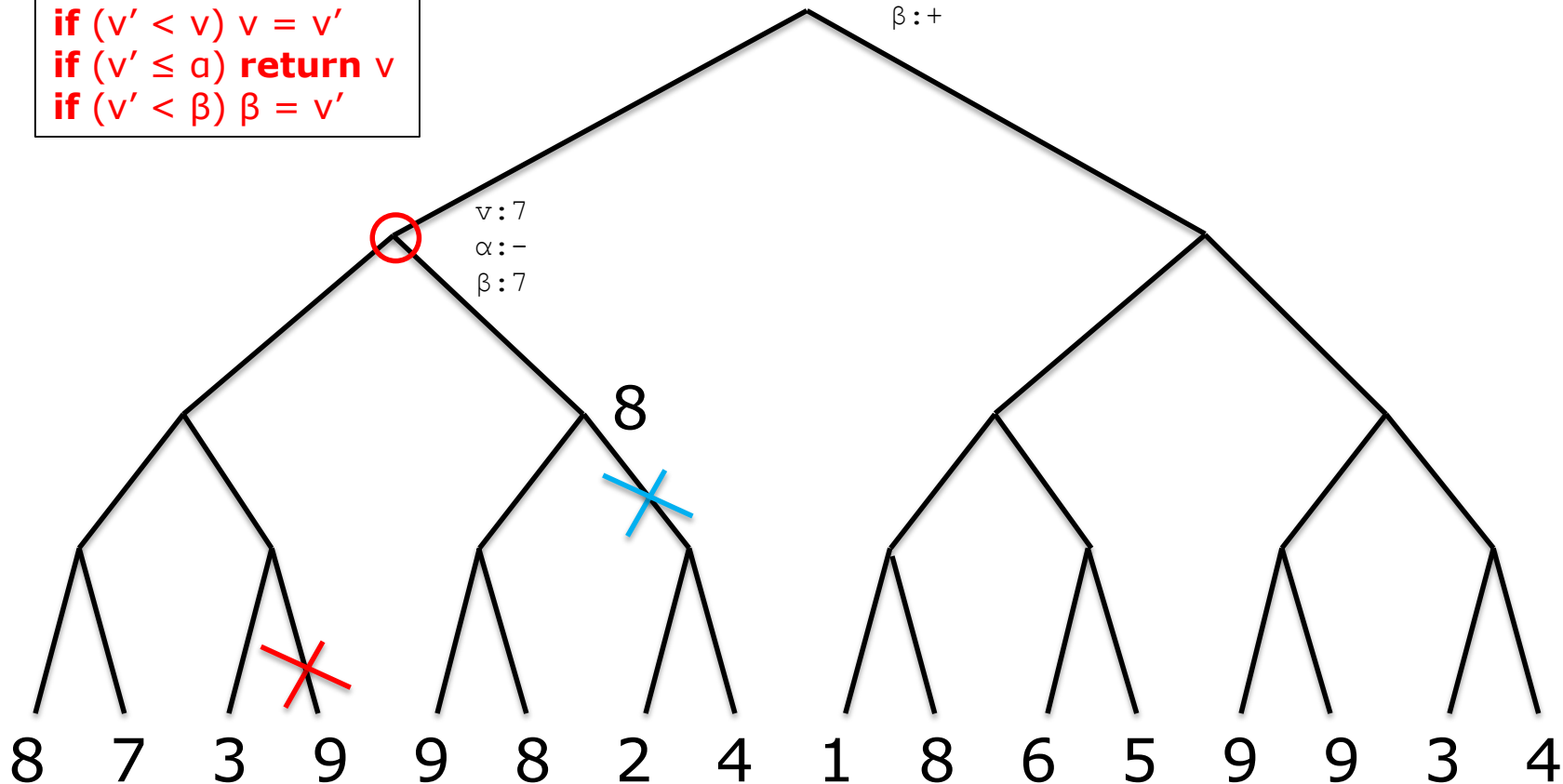
```

if ( $v' < v$ )  $v = v'$ 
if ( $v' \leq \alpha$ ) return  $v$ 
if ( $v' < \beta$ )  $\beta = v'$ 

```

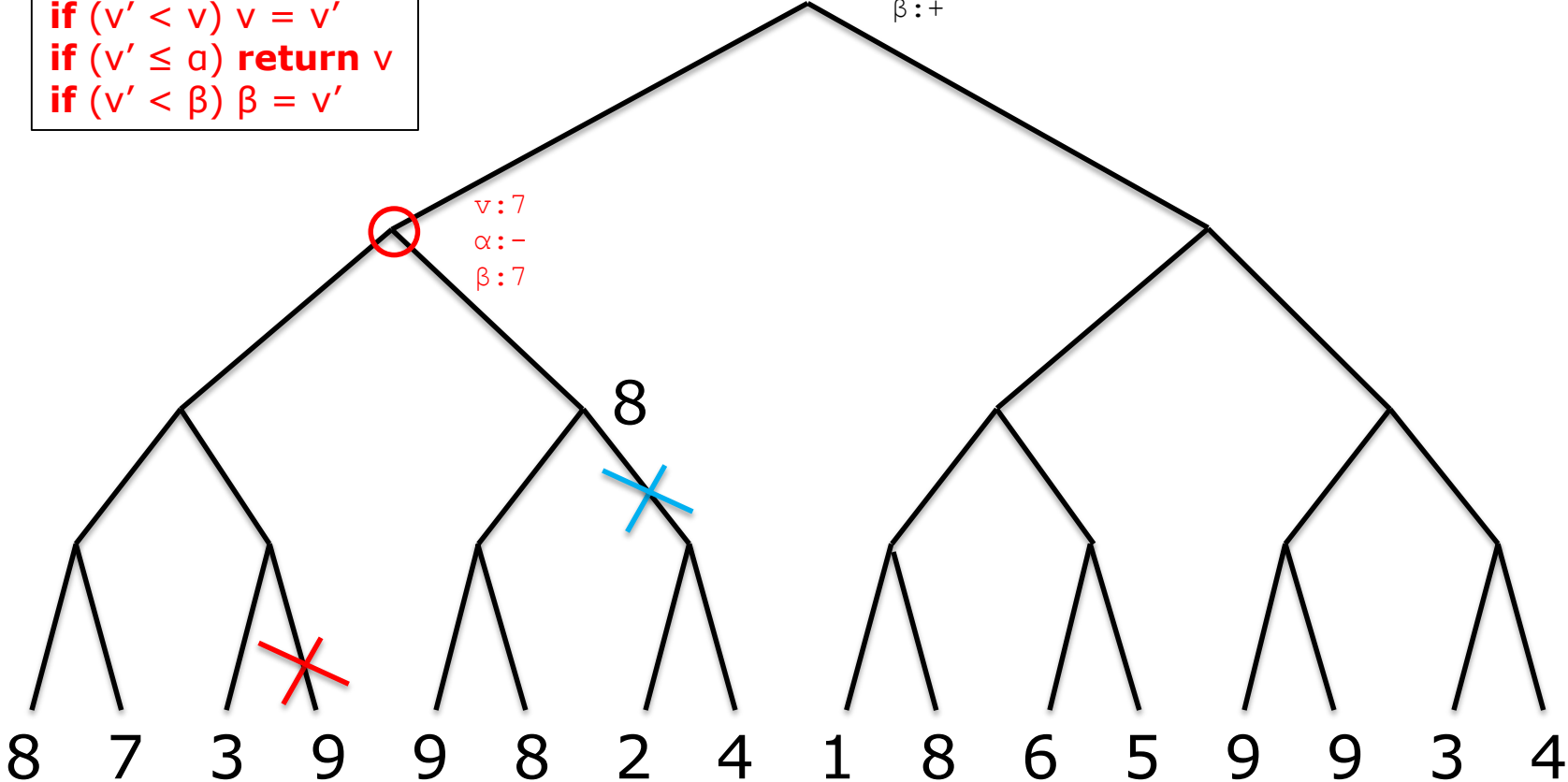
$$\begin{array}{l} \nu : - \\ \alpha : - \\ \beta : + \end{array}$$
$$\begin{aligned} v &: 7 \\ \alpha &: - \\ \beta &: 7 \end{aligned}$$

8



MinValue
if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

$v:-$
 $\alpha:-$
 $\beta:+$

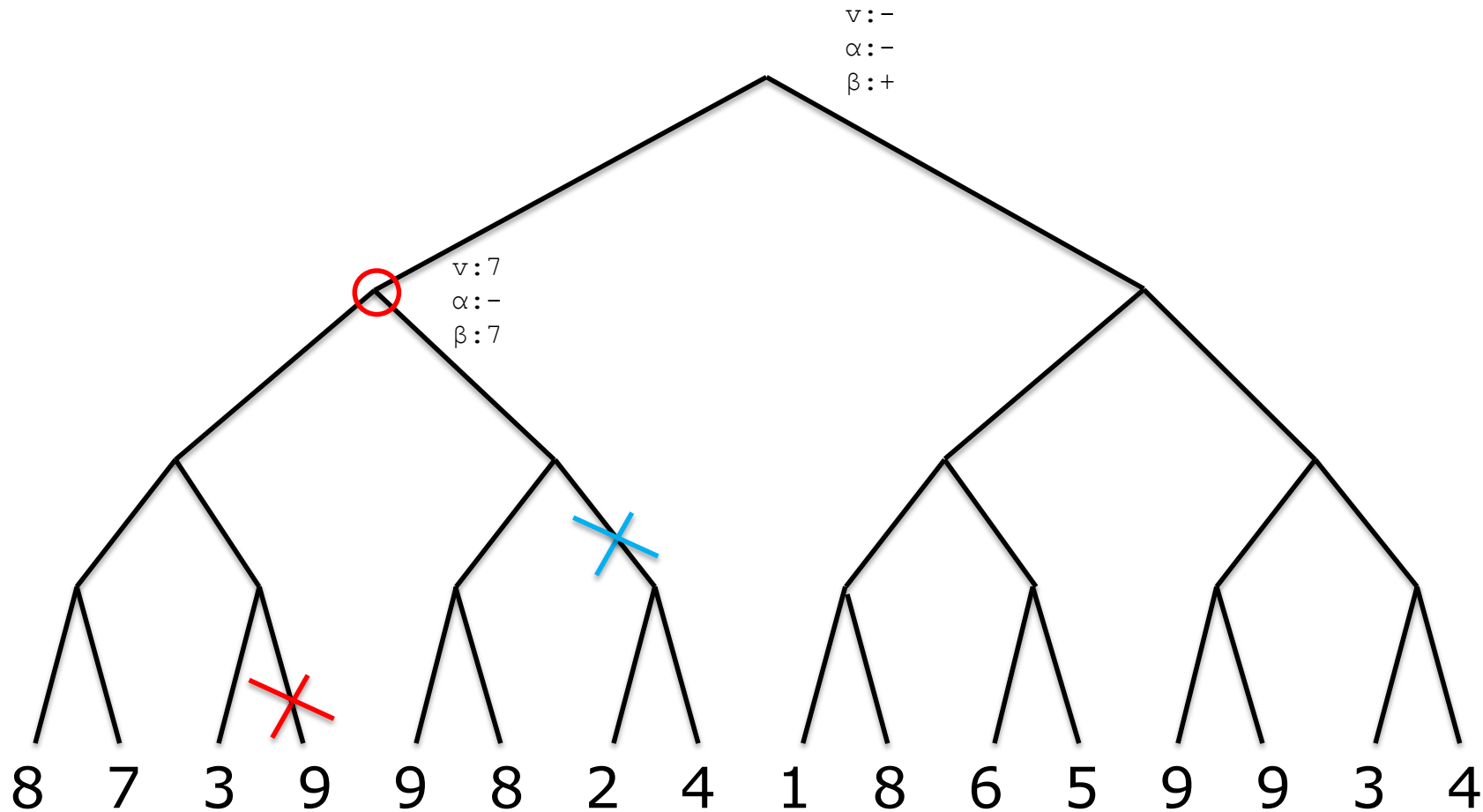


Max

Min

Max

Min



Max

Min

Max

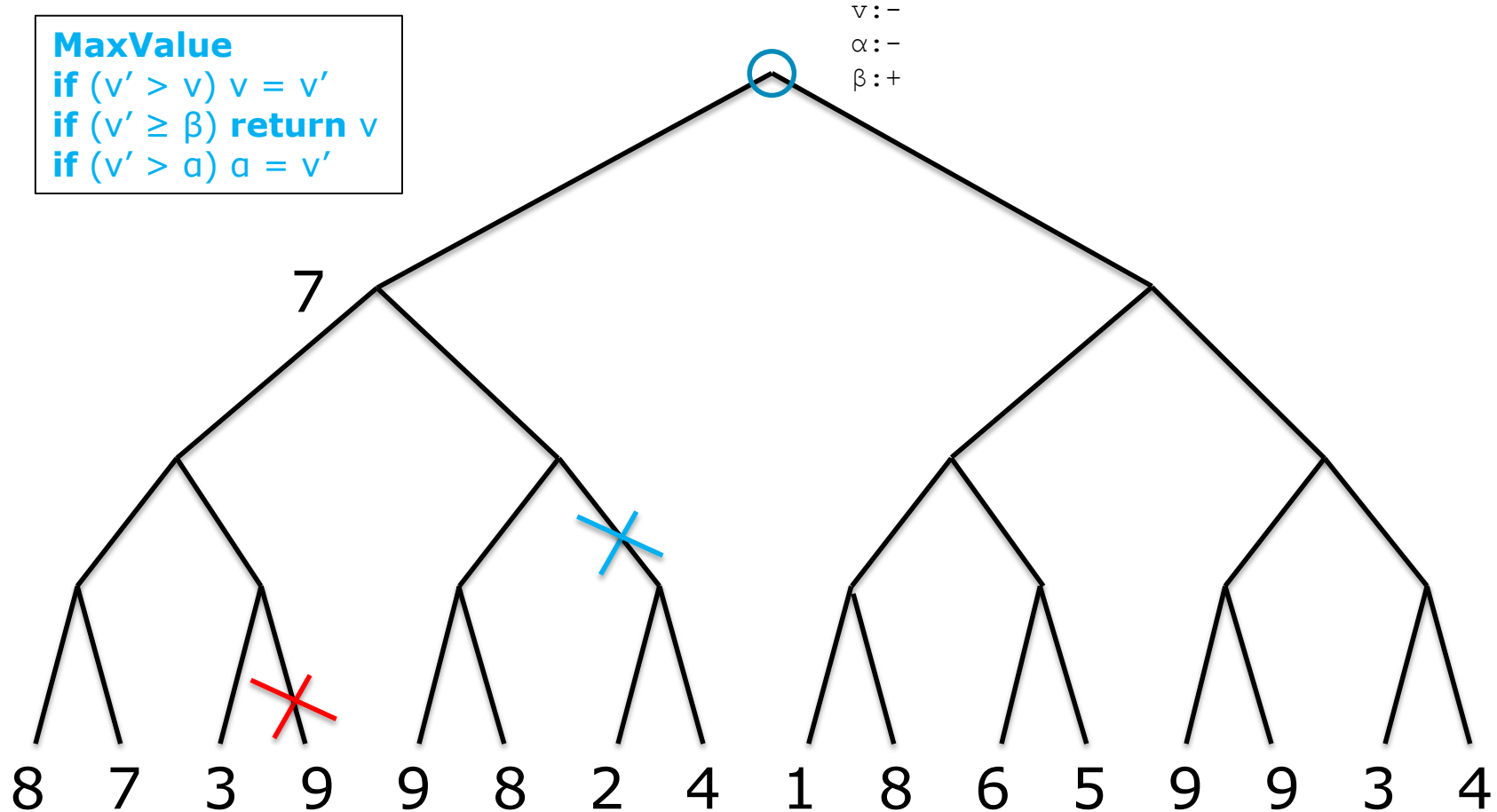
Min

MaxValue

```

if ( $v' > v$ )  $v = v'$ 
if ( $v' \geq \beta$ ) return  $v$ 
if ( $v' > a$ )  $a = v'$ 

```

$$V: -$$
$$\alpha :-$$
 $\beta : +$ 

Max

MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

$v:7$

$\alpha:7$

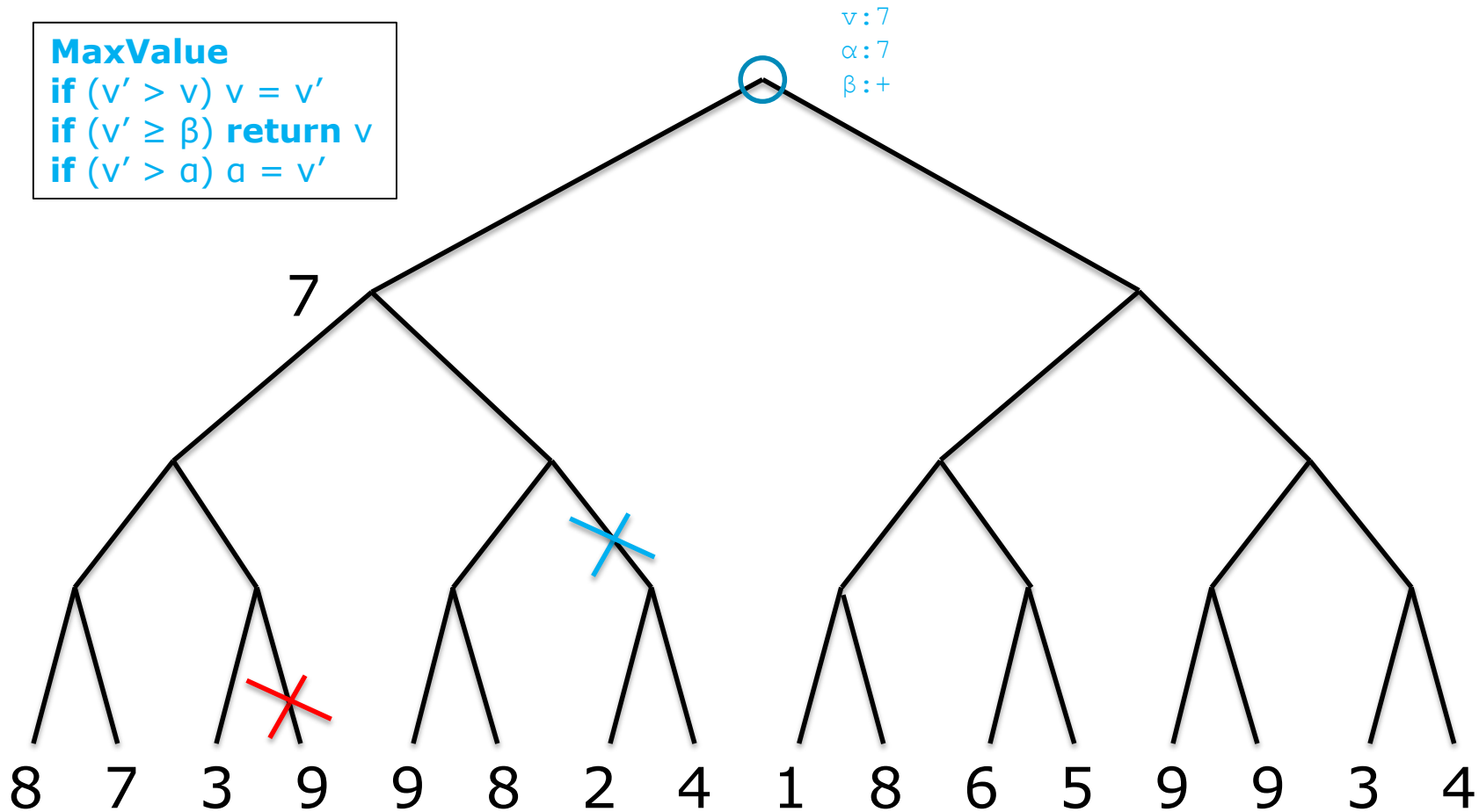
$\beta:+$

Min

7

Max

Min

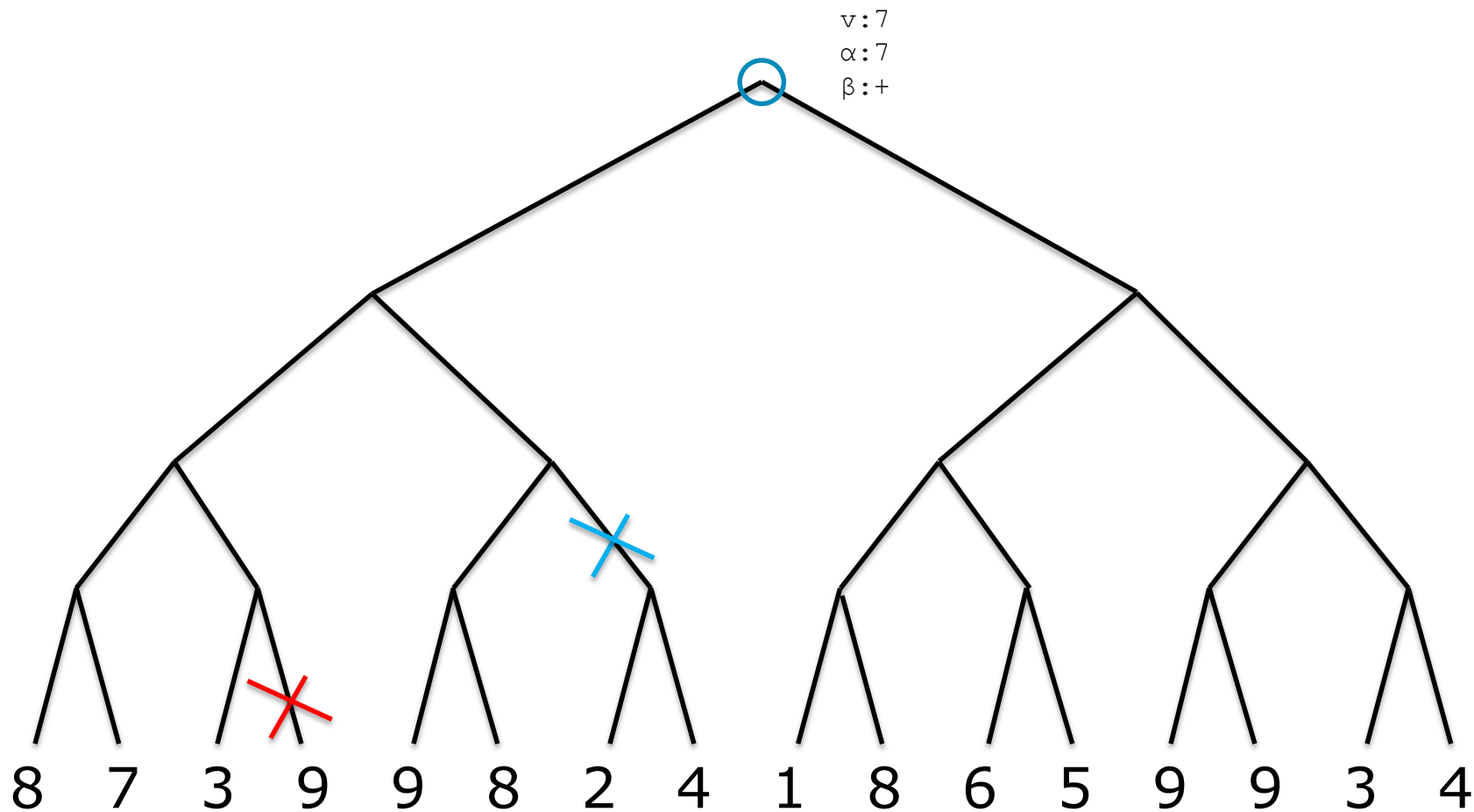


Max

Min

Max

Min

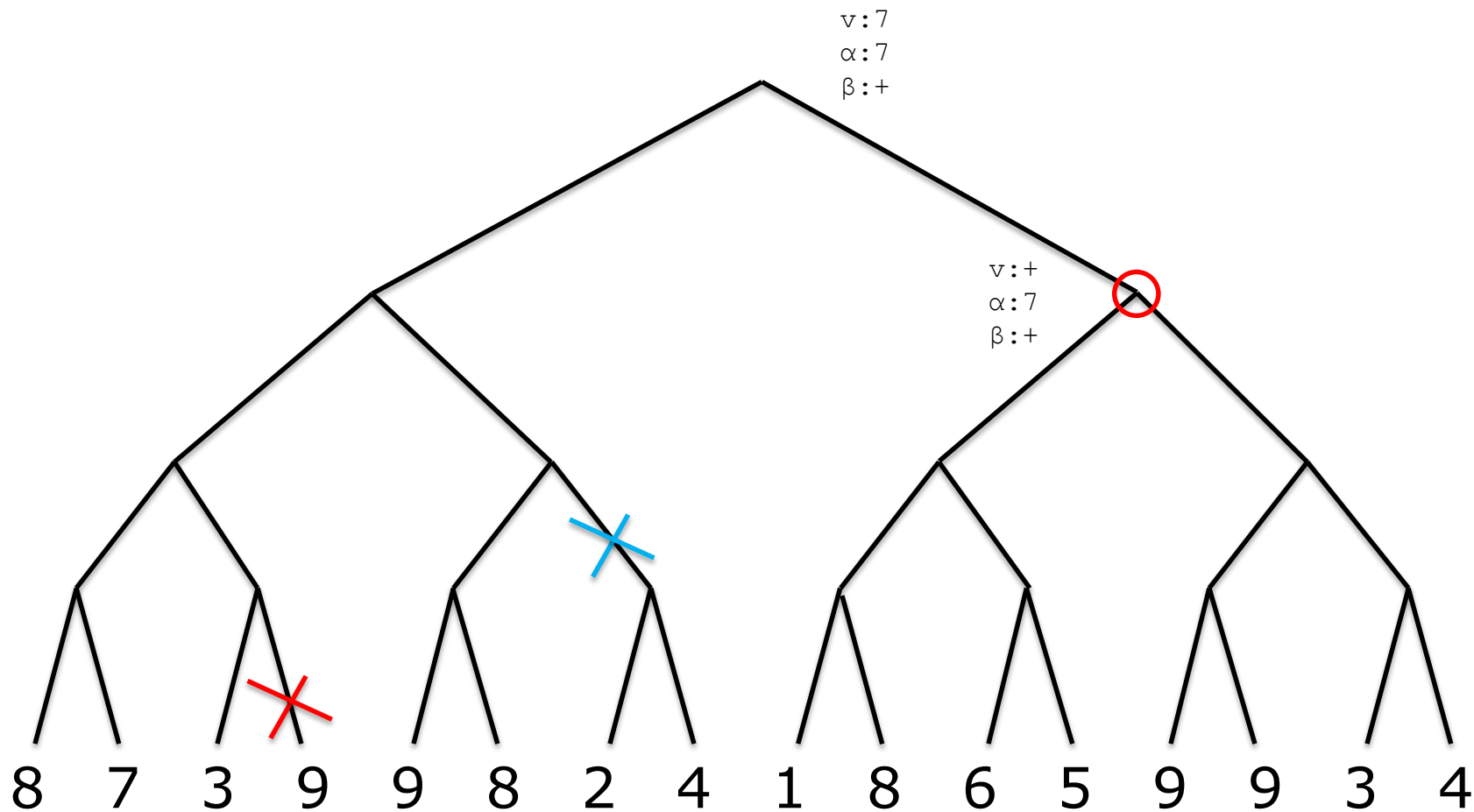


Max

Min

Max

Min

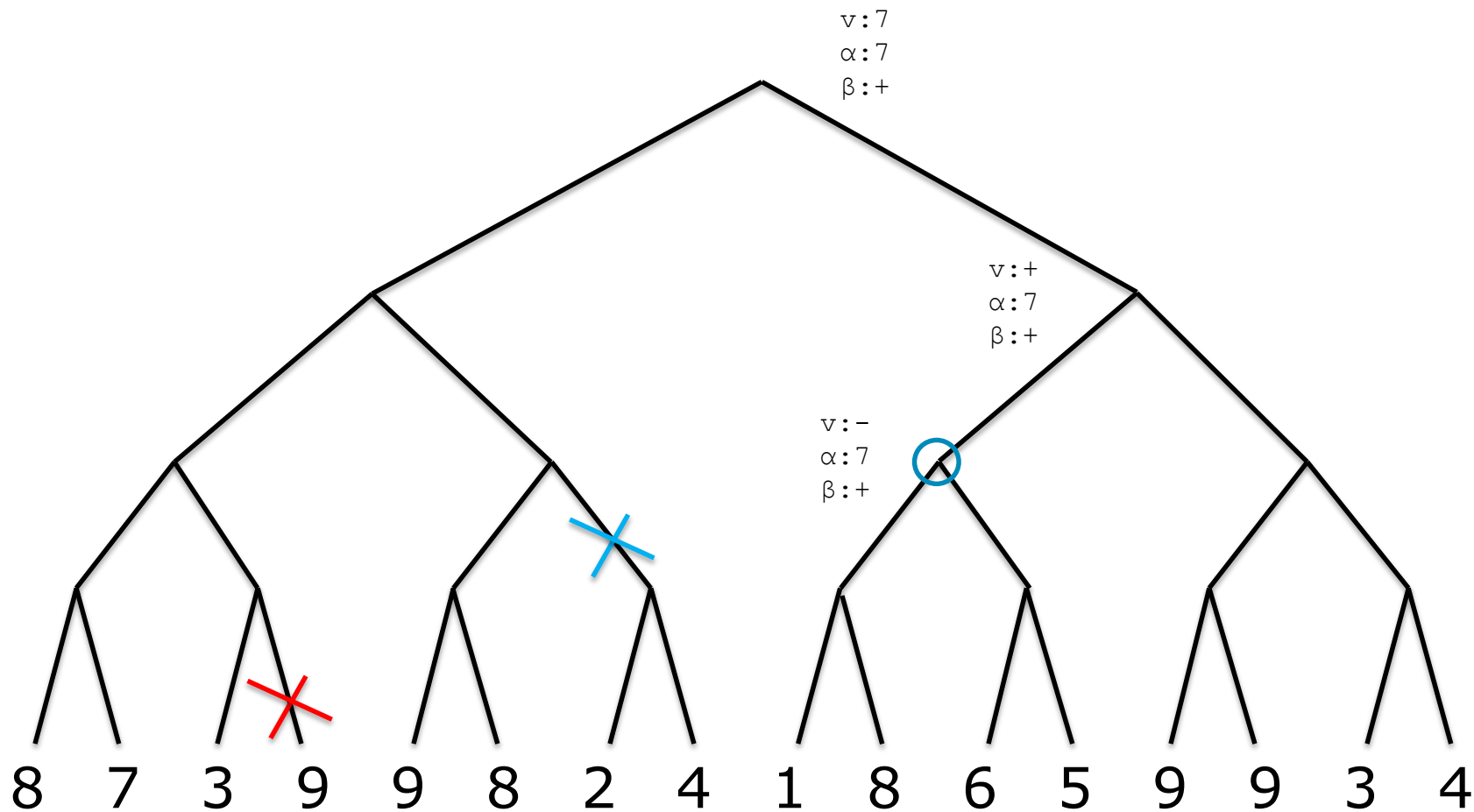


Max

Min

Max

Min

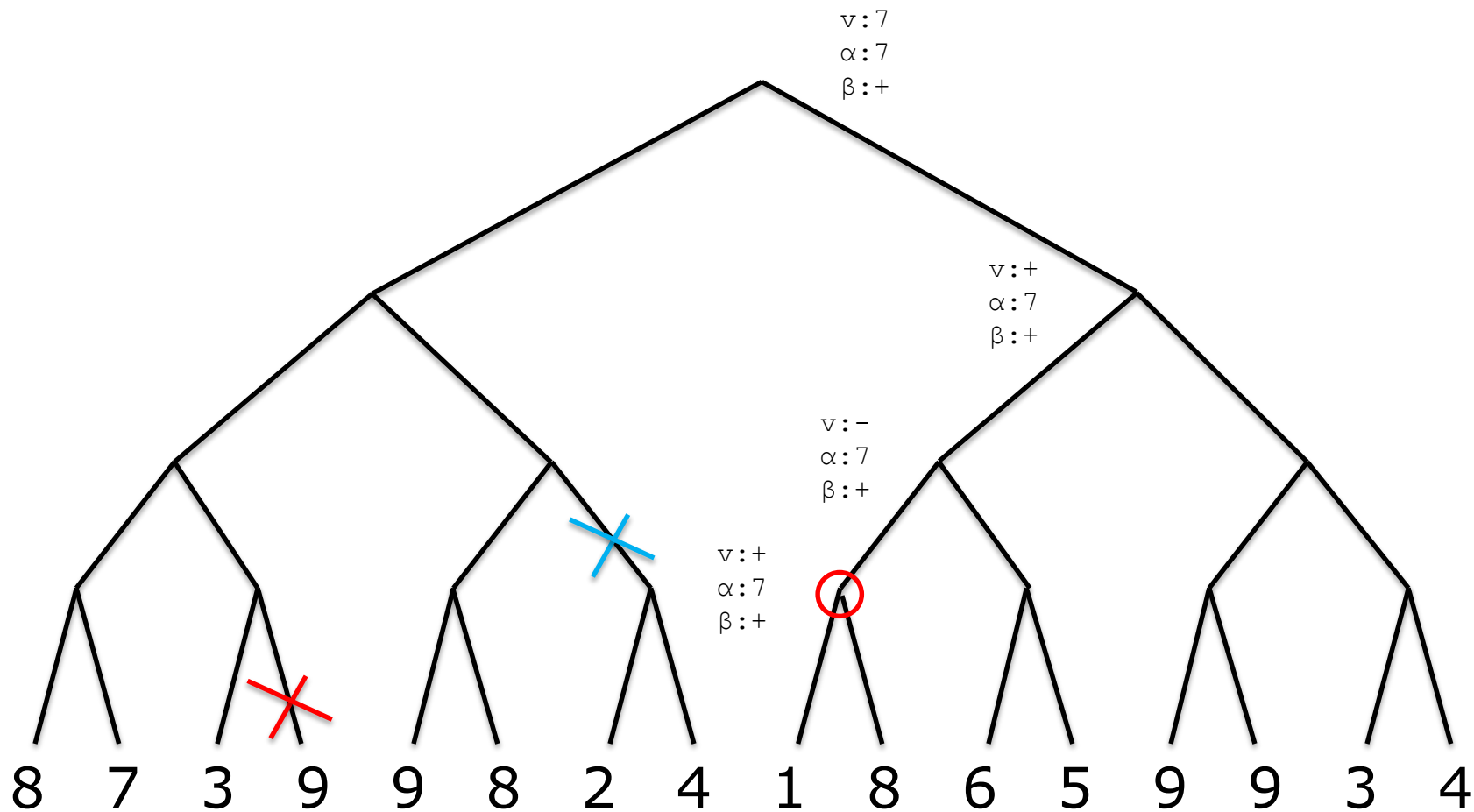


Max

Min

Max

Min




```

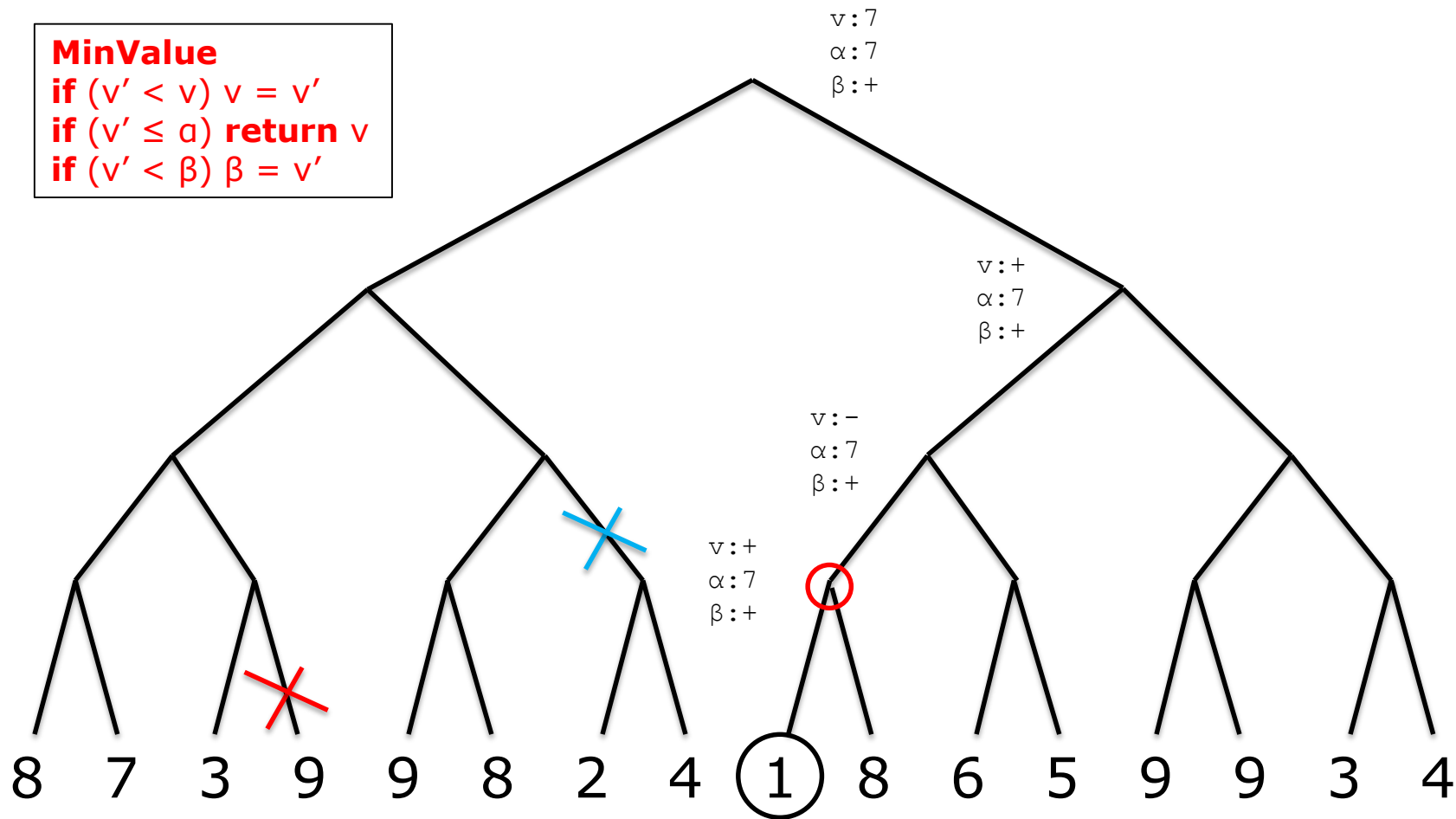
if ( $v' < v$ )  $v = v'$ 
if ( $v' \leq \alpha$ ) return  $v$ 
if ( $v' < \beta$ )  $\beta = v'$ 

```

Min

Max

Min



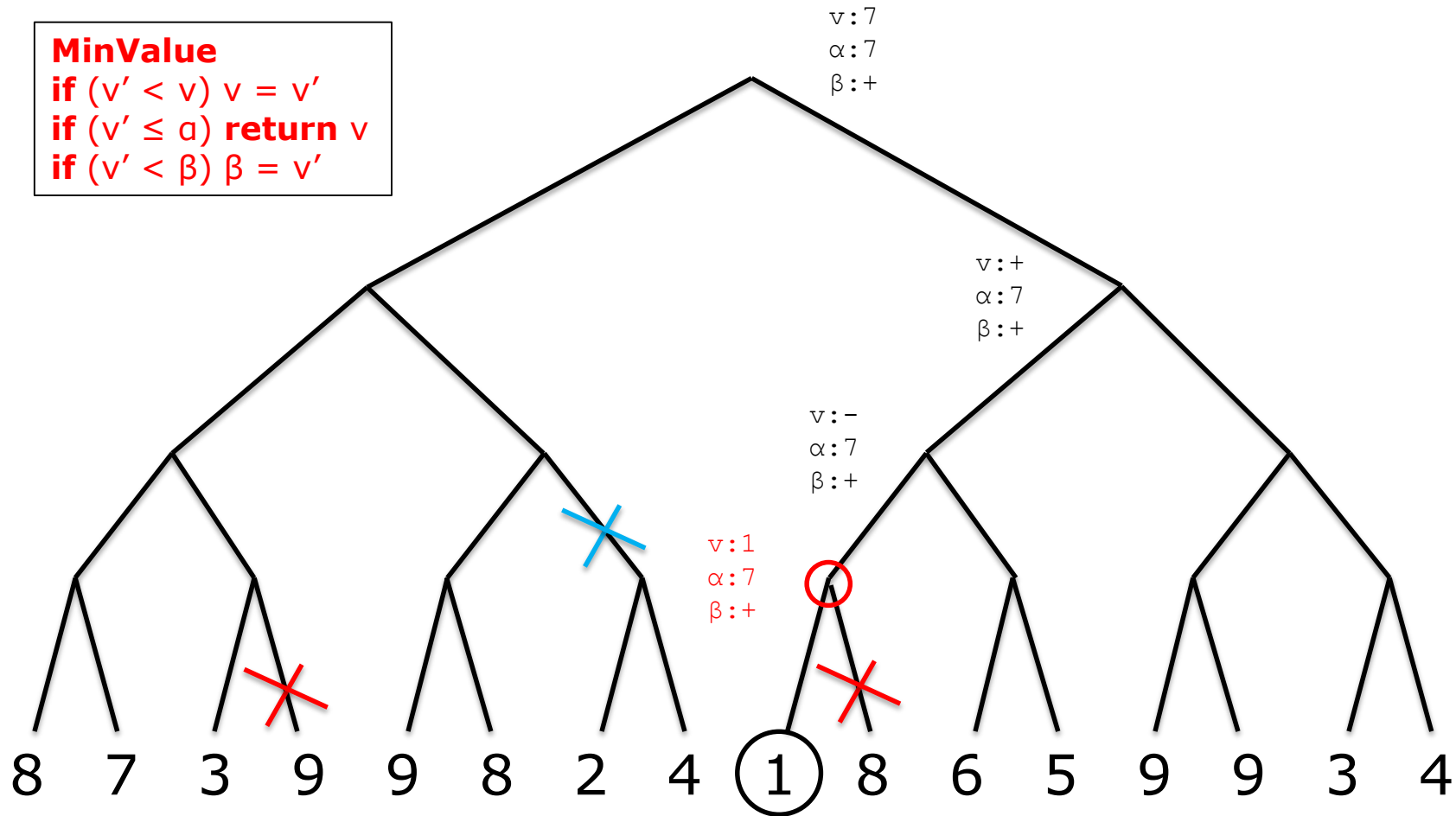
MinValue
if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

Max

Min

Max

Min



Max

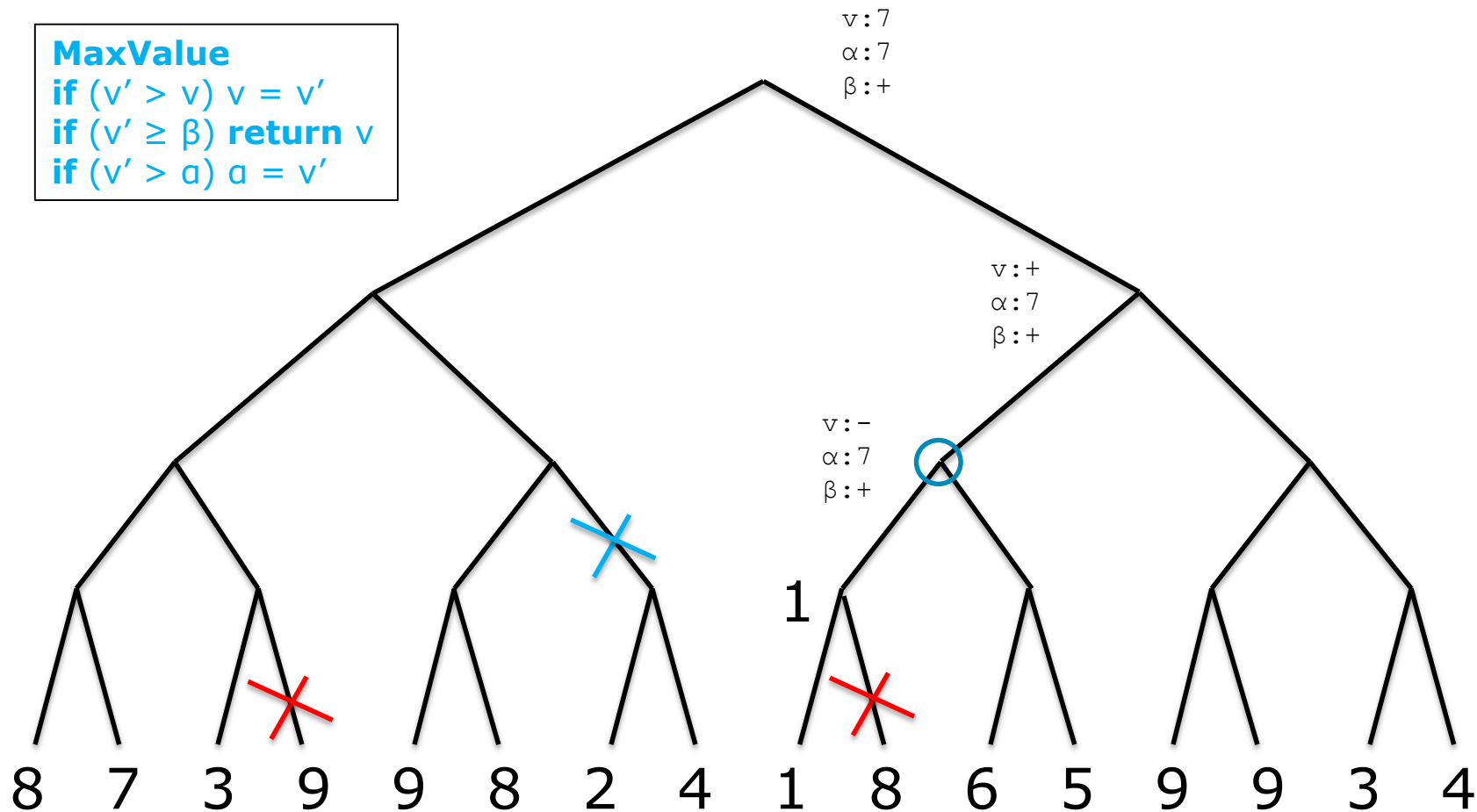
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



Max

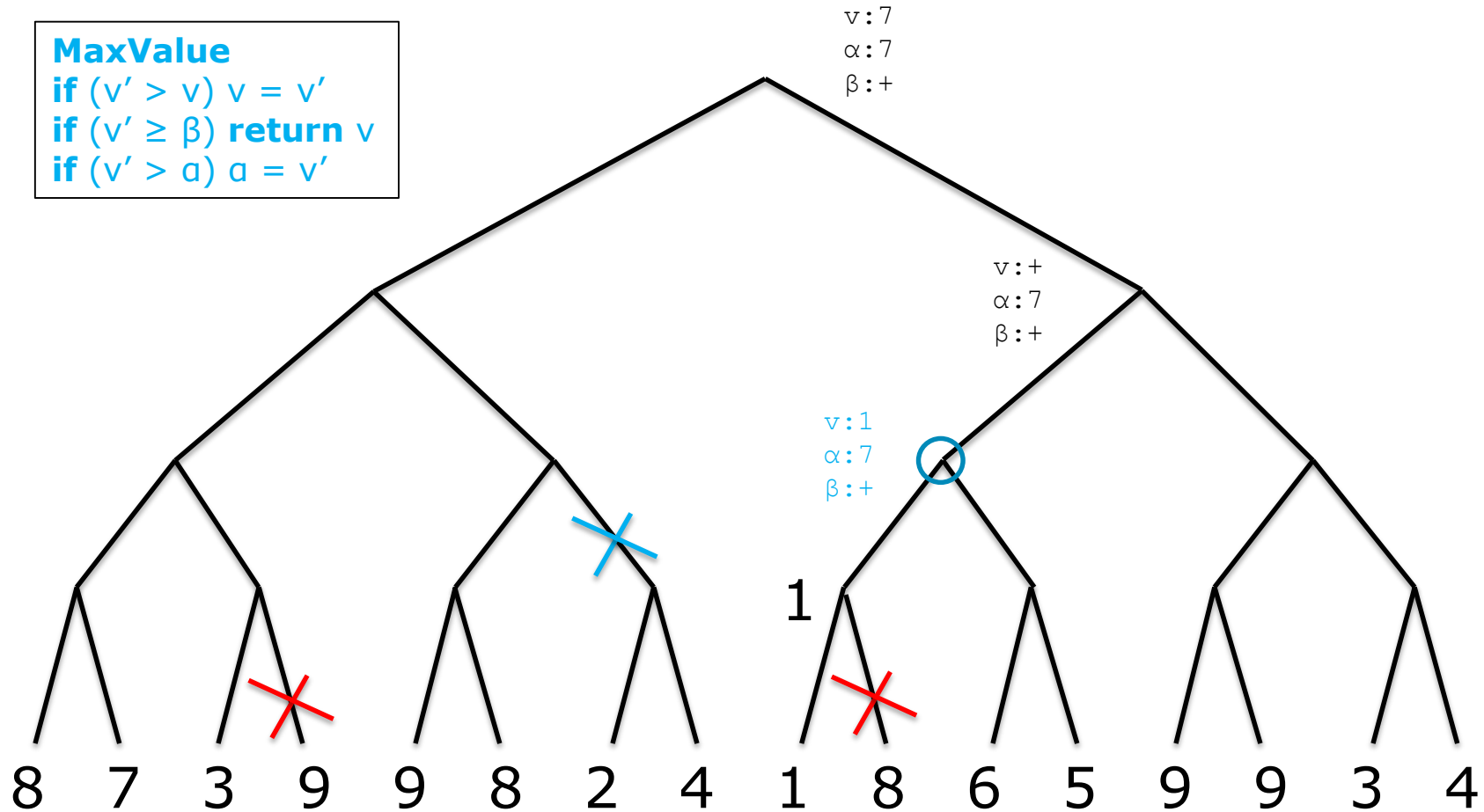
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min

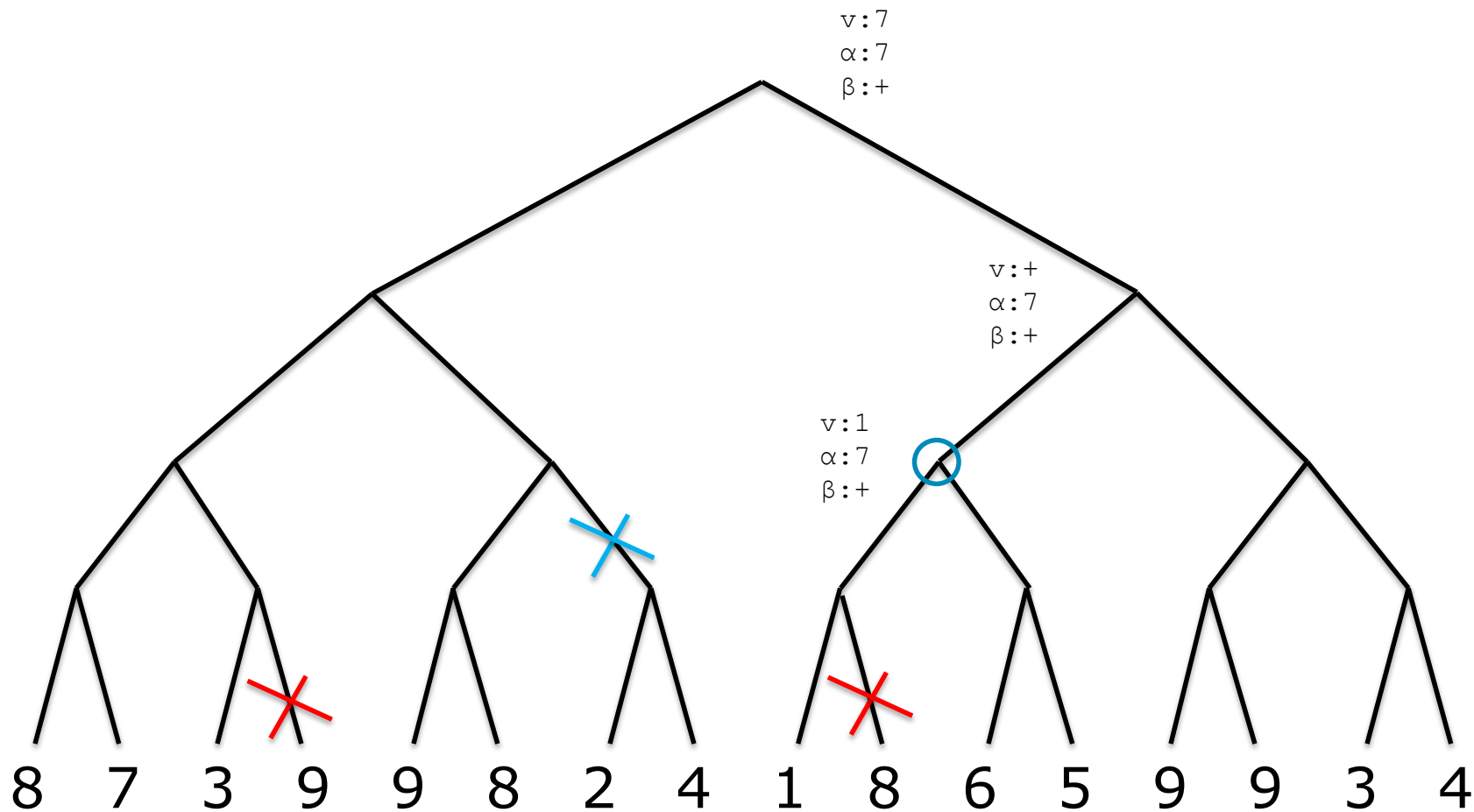


Max

Min

Max

Min

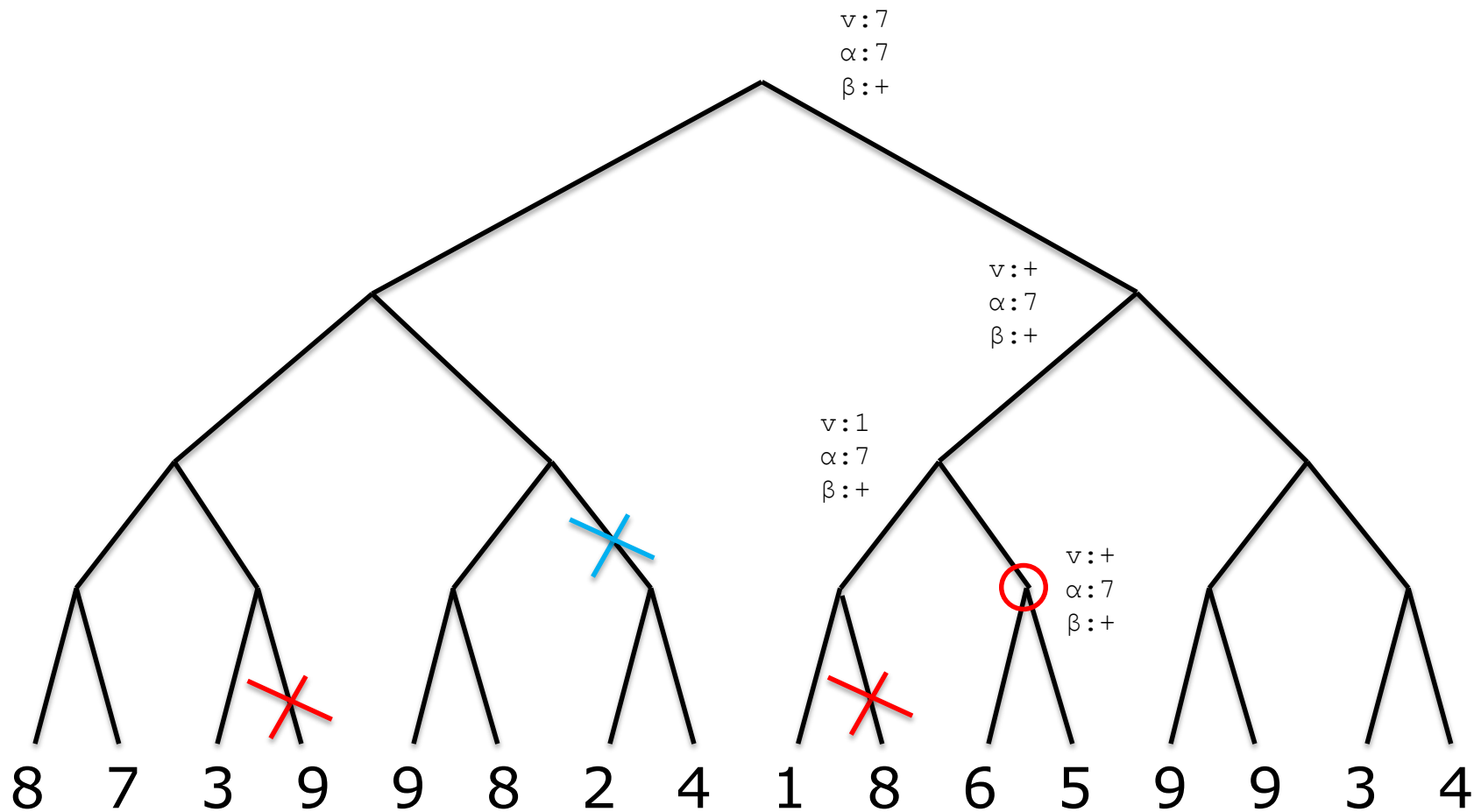


Max

Min

Max

Min



Max

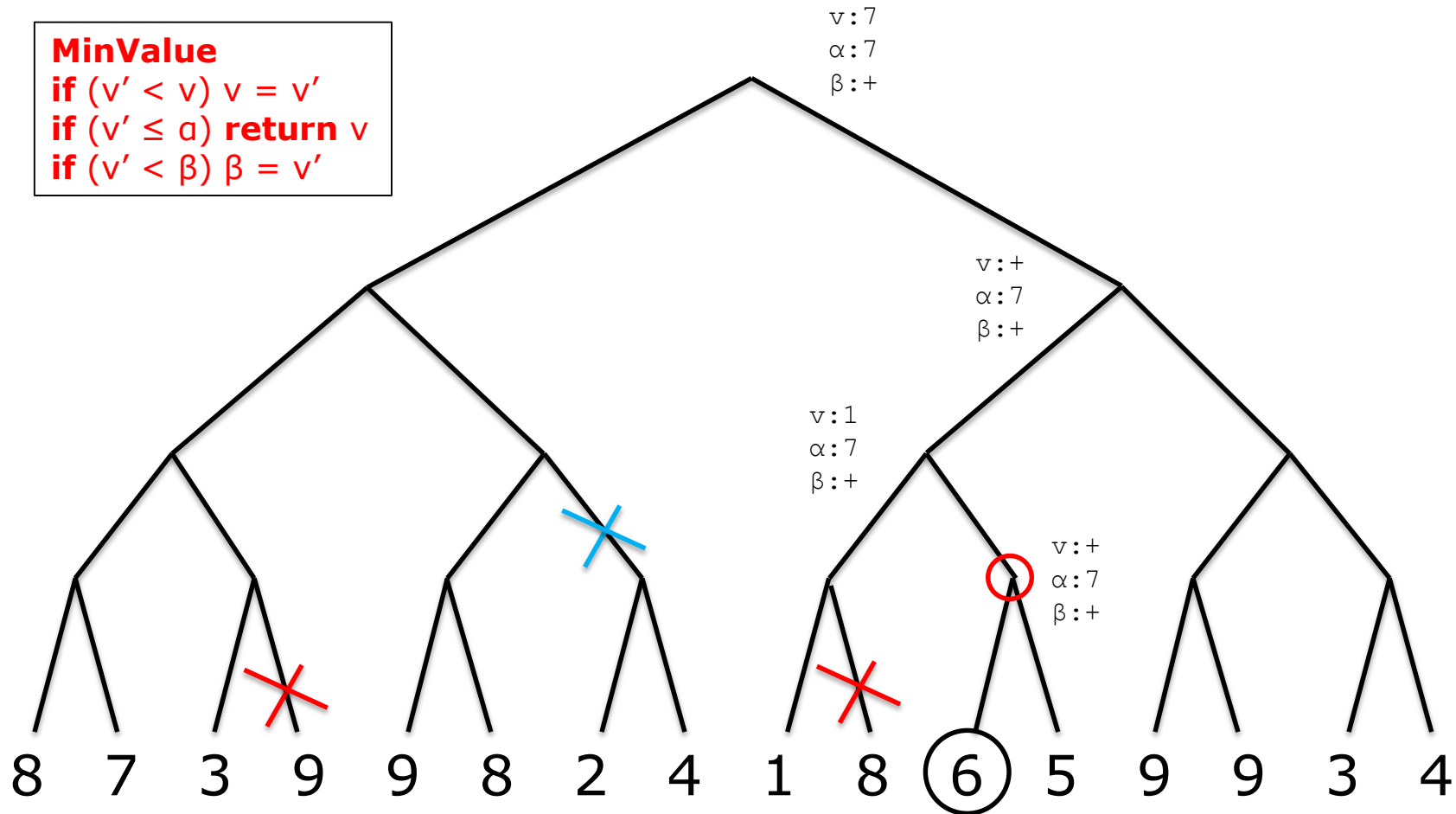
MinValue

if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

Min

Max

Min



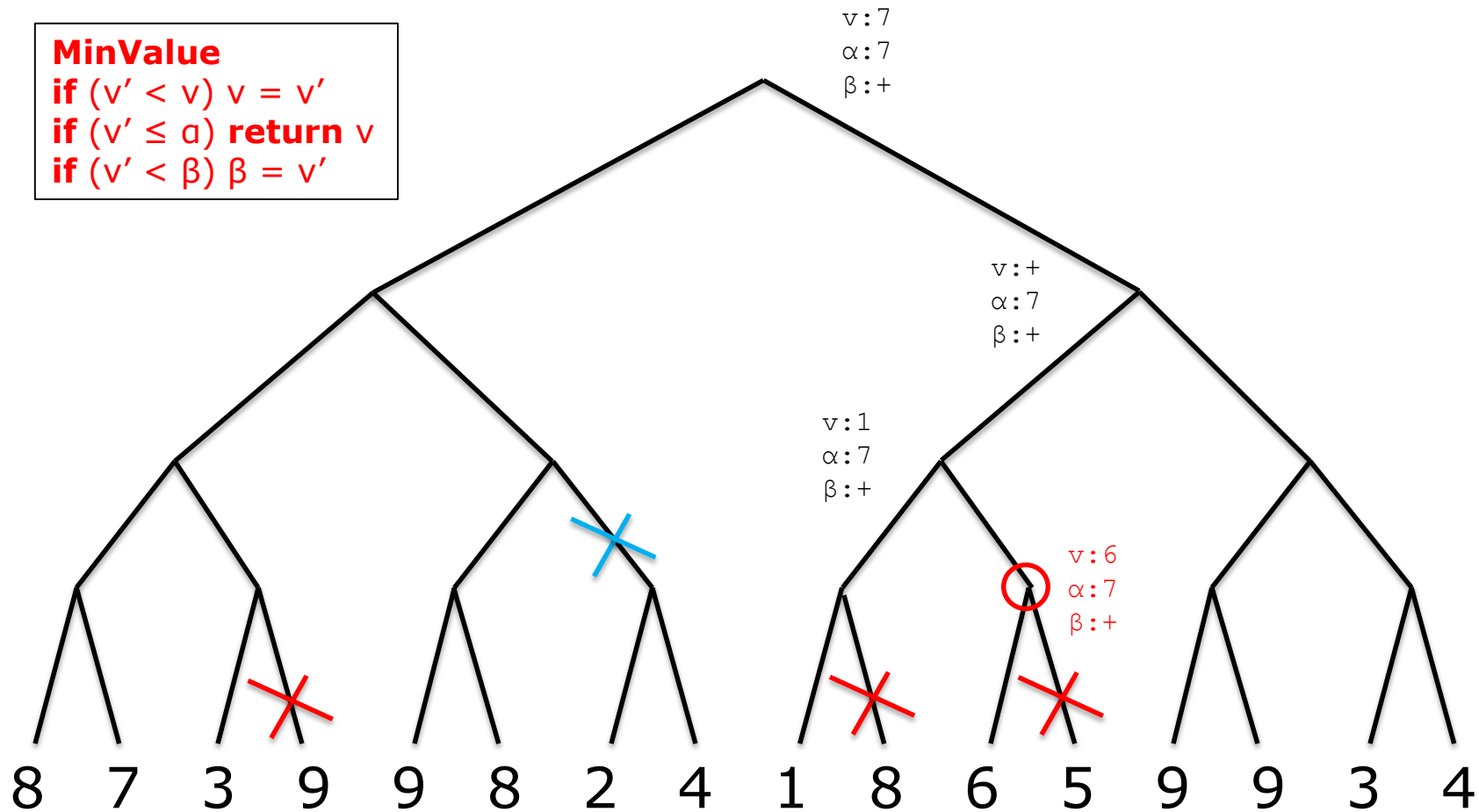
MinValue
if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

Max

Min

Max

Min



Max

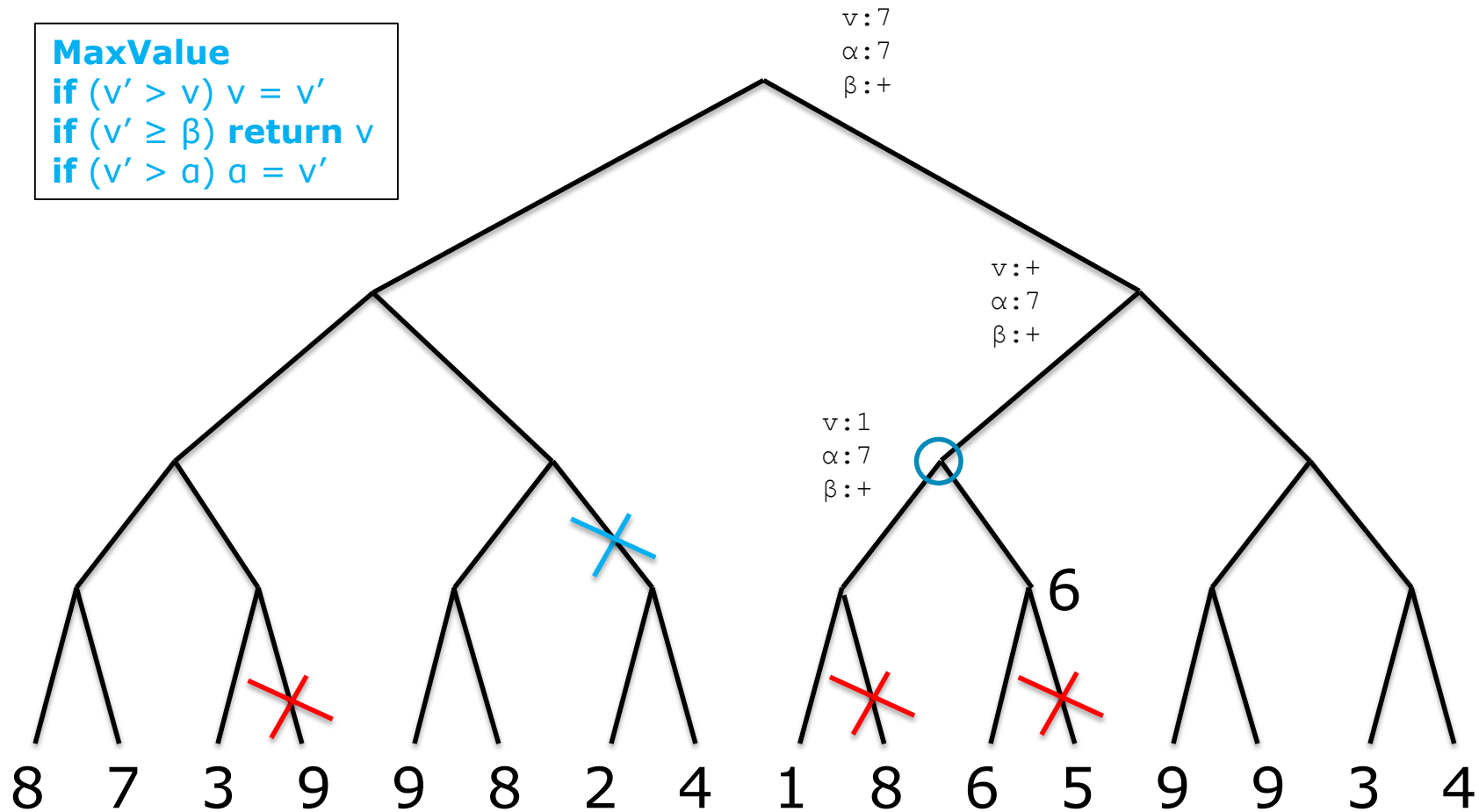
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



Max

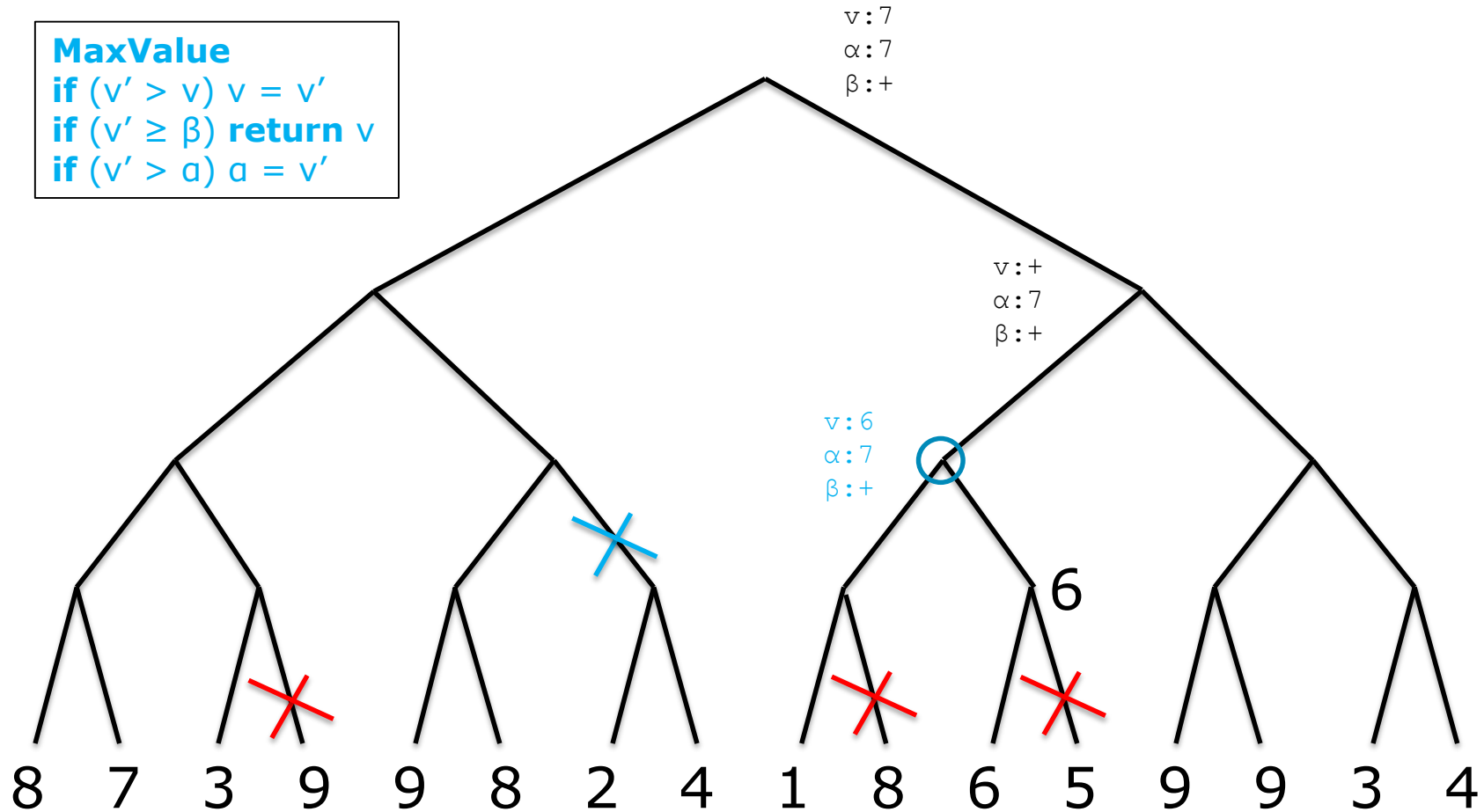
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



Max

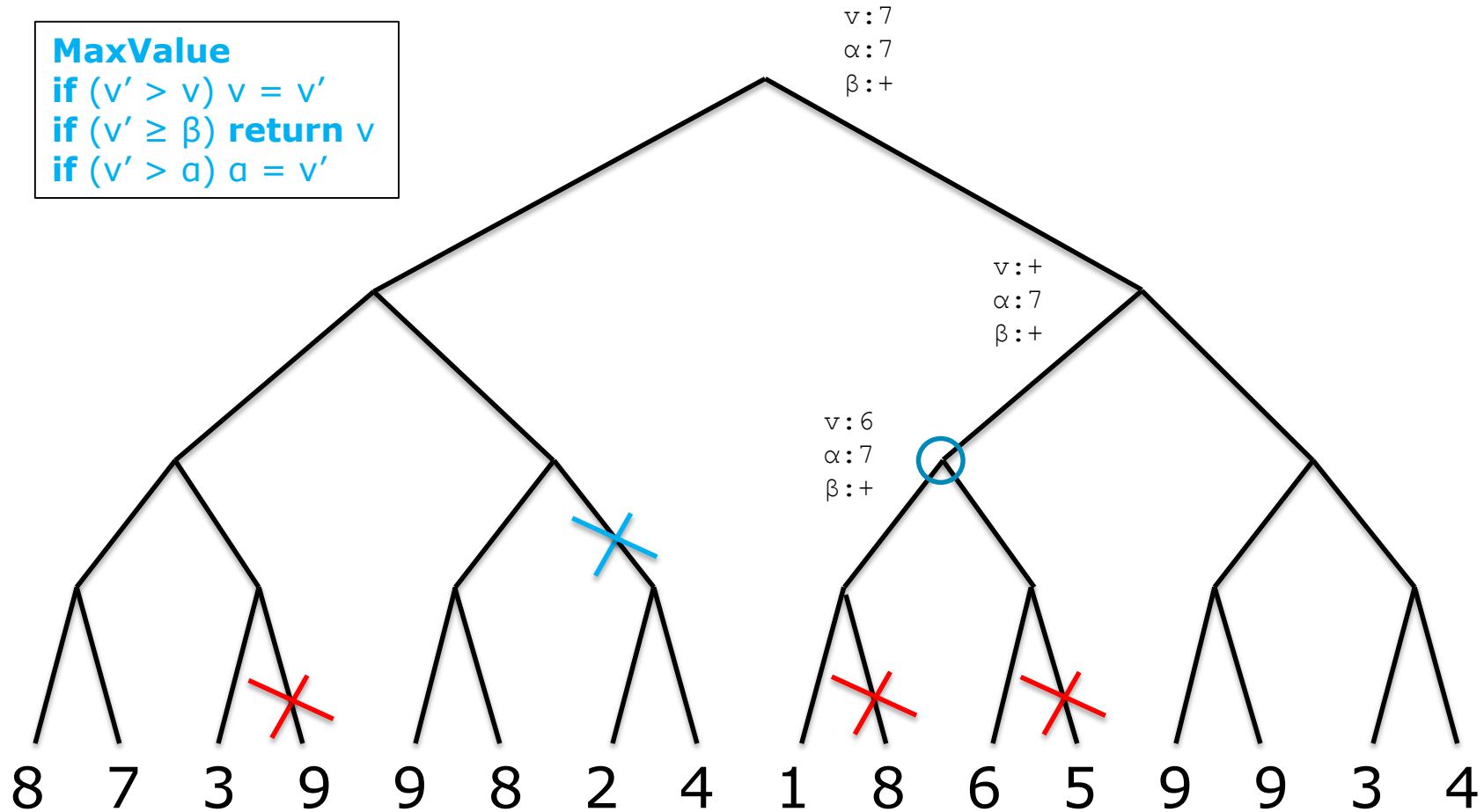
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



Min

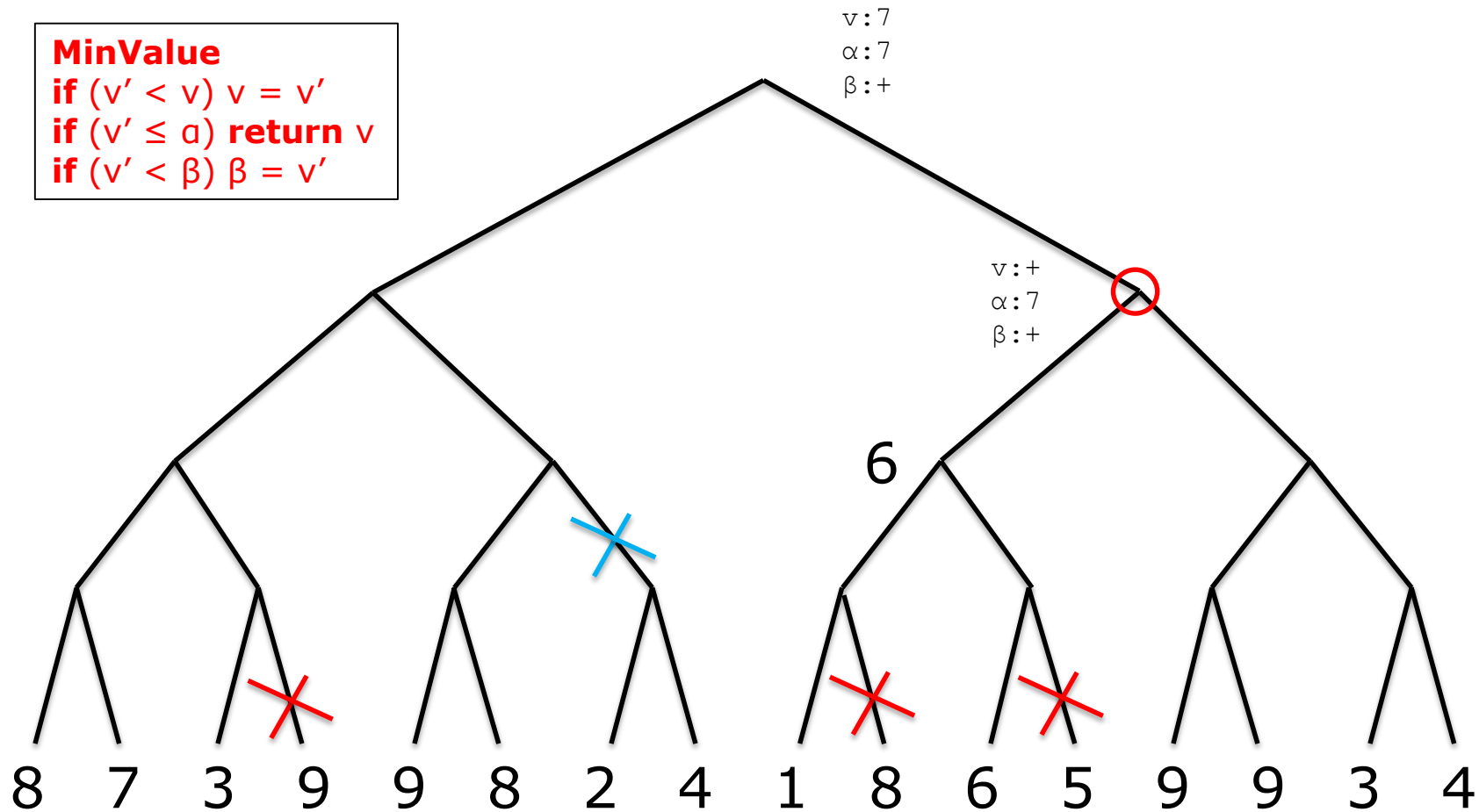
Max

Min

```

if ( $v' < v$ )  $v = v'$ 
if ( $v' \leq \alpha$ ) return  $v$ 
if ( $v' < \beta$ )  $\beta = v'$ 

```



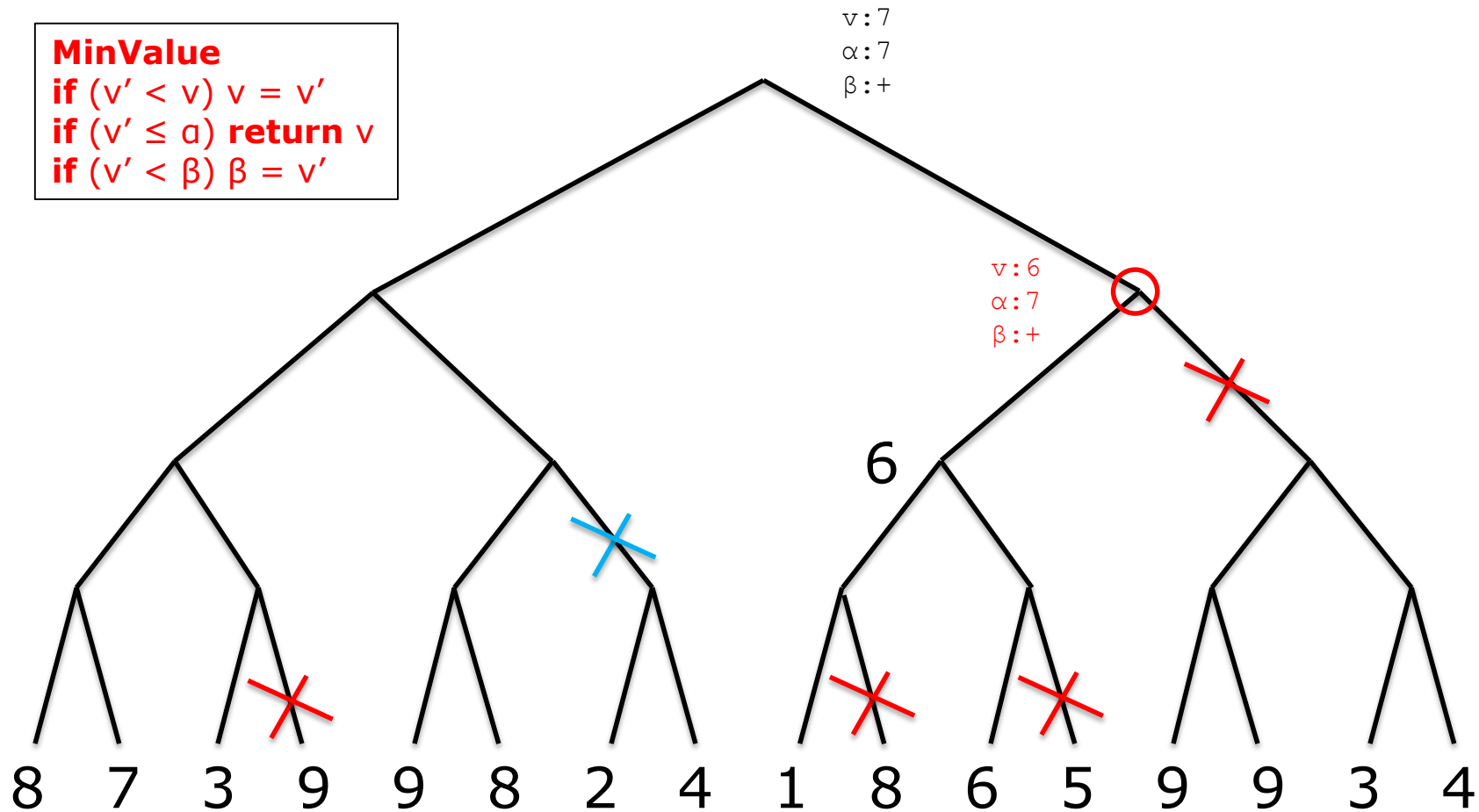
Max

MinValue
if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

Min

Max

Min



Max

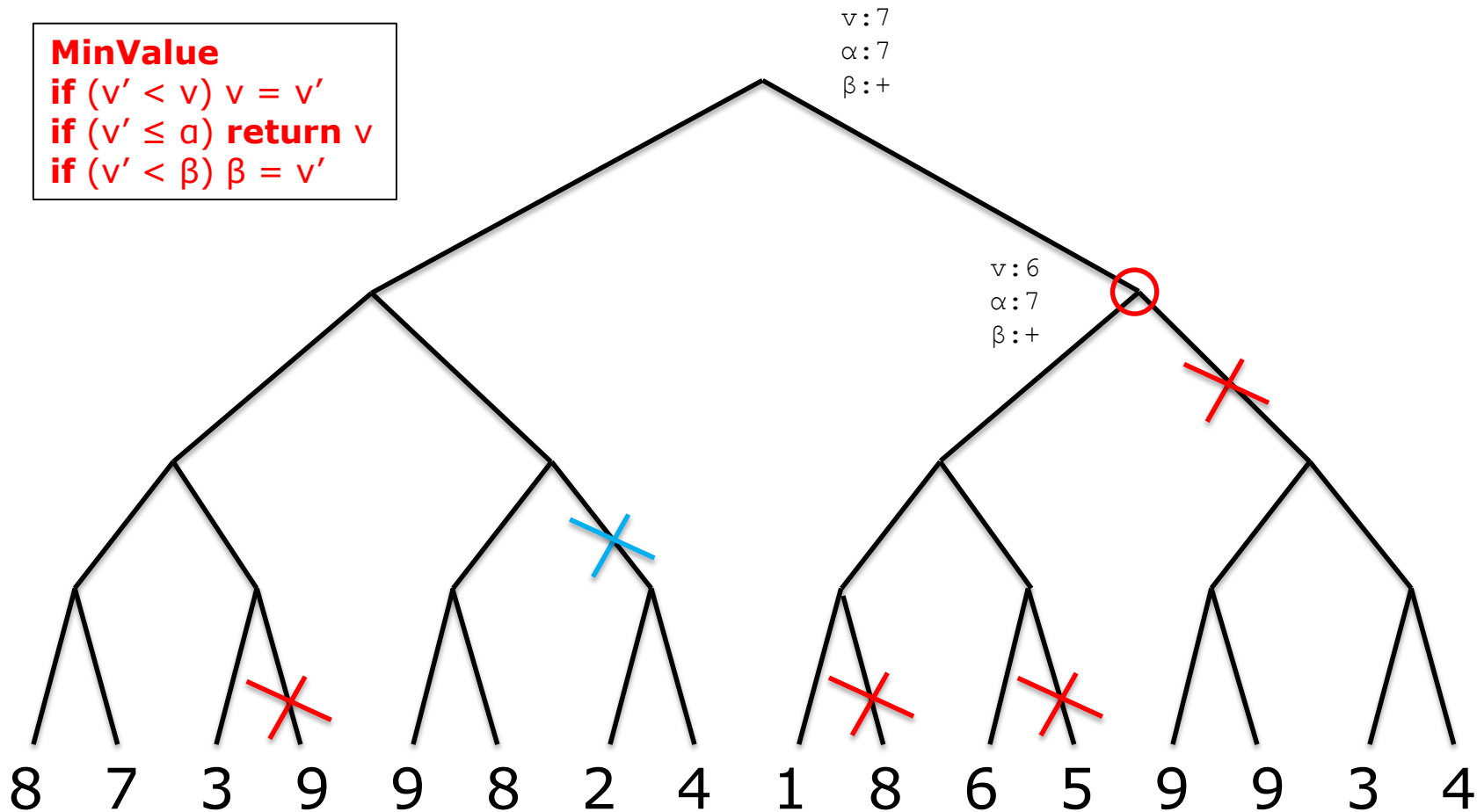
MinValue

if ($v' < v$) $v = v'$
if ($v' \leq \alpha$) **return** v
if ($v' < \beta$) $\beta = v'$

Min

Max

Min



Max

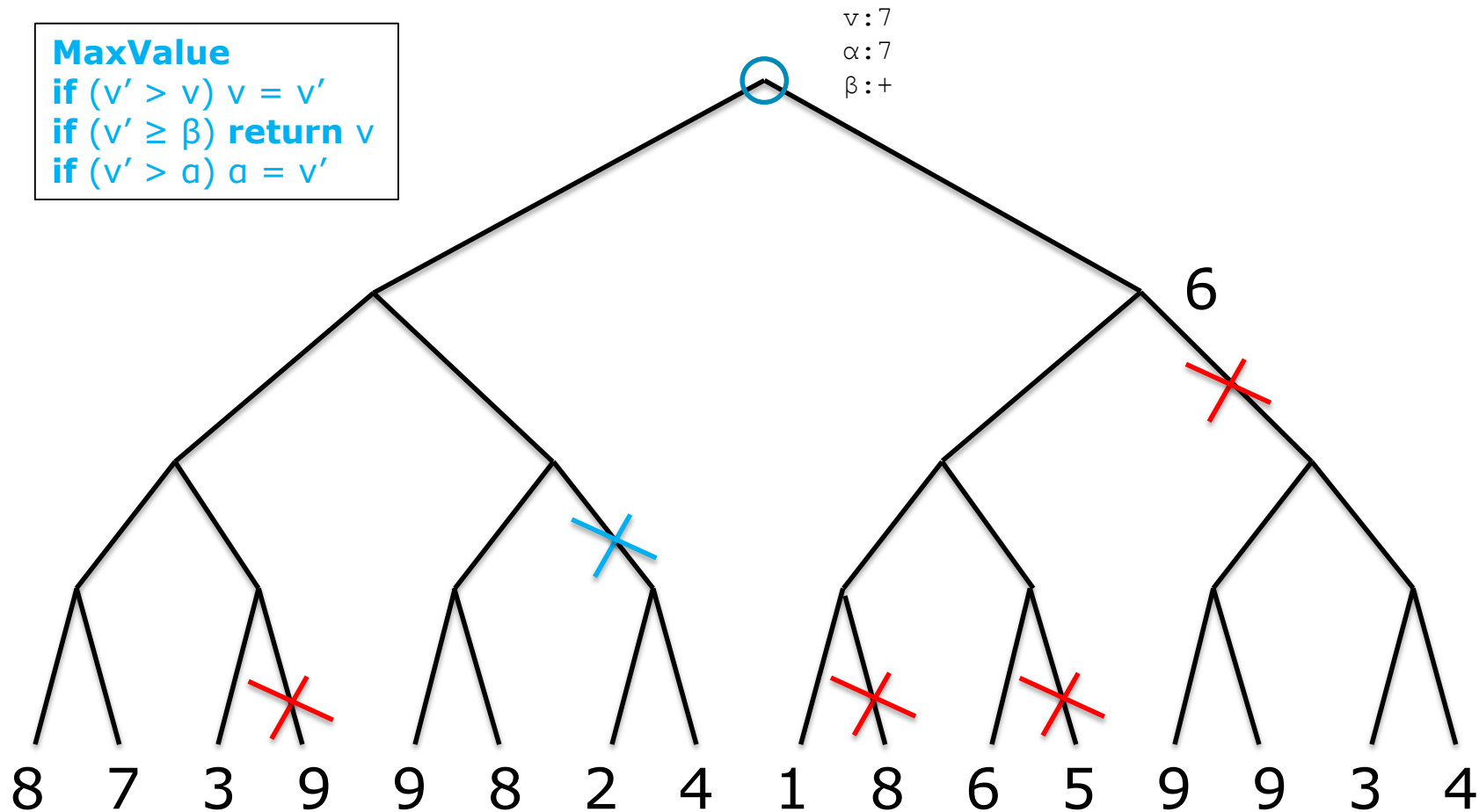
MaxValue

```
if ( $v' > v$ )  $v = v'$   
if ( $v' \geq \beta$ ) return  $v$   
if ( $v' > \alpha$ )  $\alpha = v'$ 
```

Min

Max

Min



Max

MaxValue

```

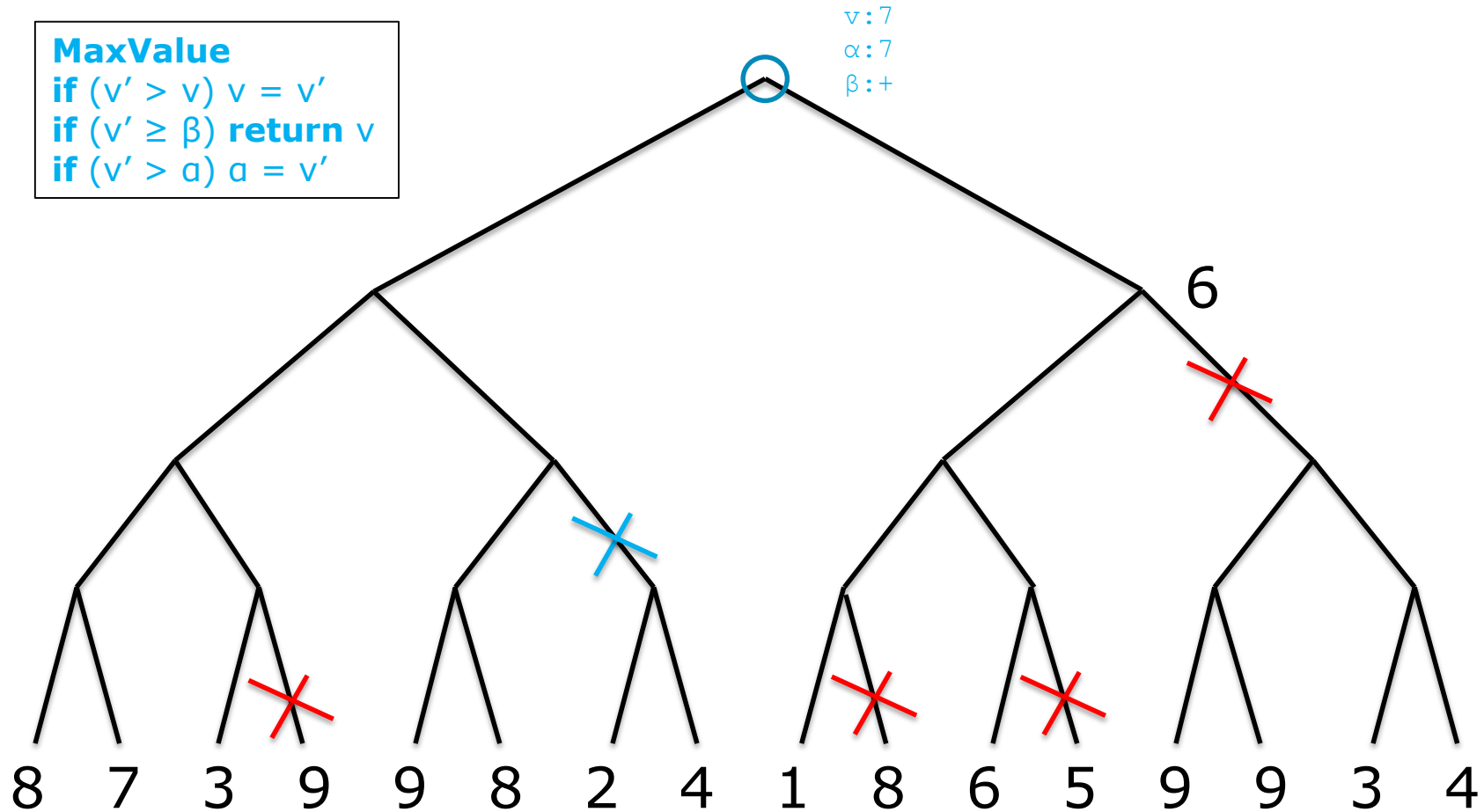
if ( $v' > v$ )  $v = v'$ 
if ( $v' \geq \beta$ ) return  $v$ 
if ( $v' > a$ )  $a = v'$ 

```

Min

Max

Min

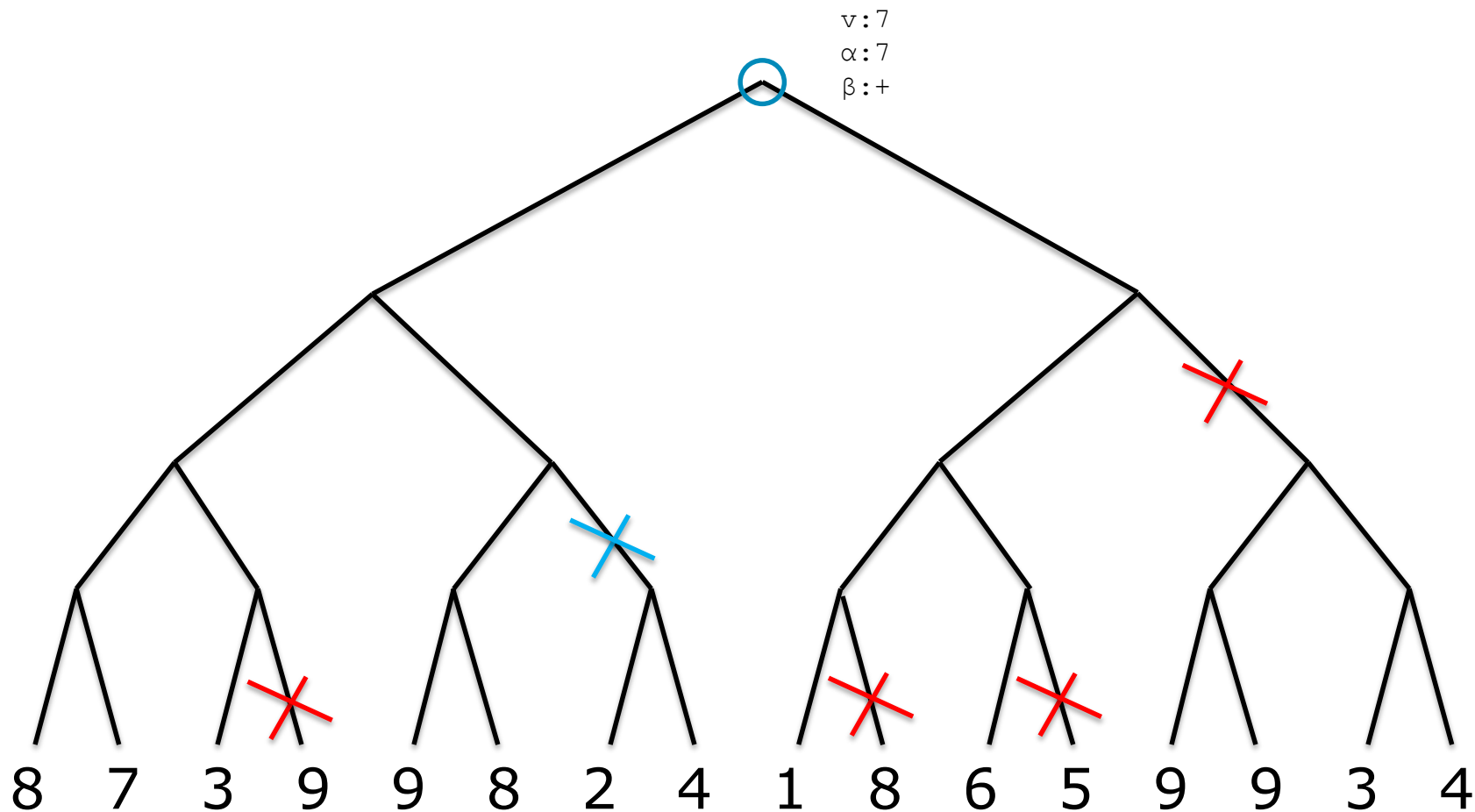


Max

Min

Max

Min



Max

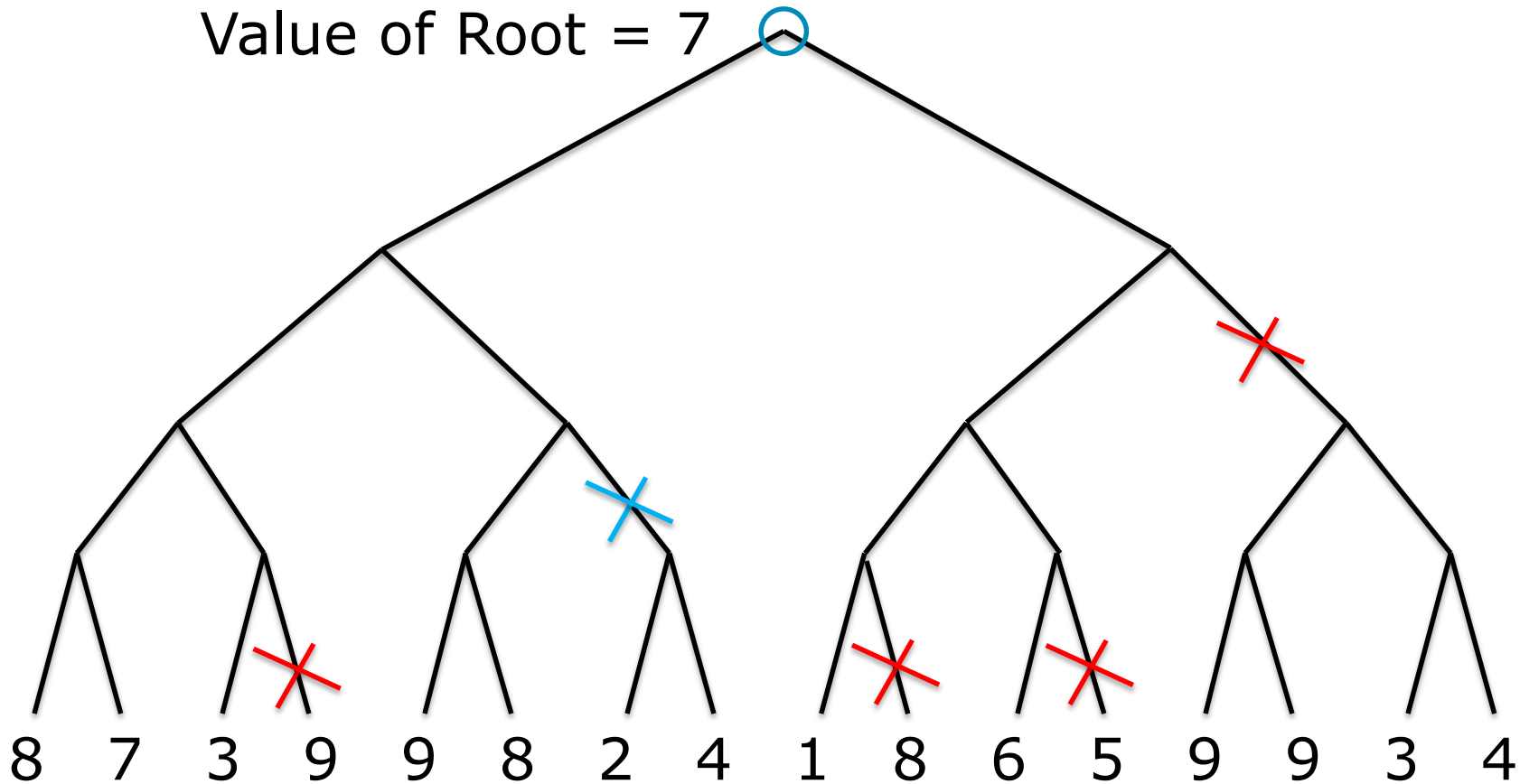
Value of Root = 7



Min

Max

Min



Computational Savings

- MiniMax Tree Search
 - Nodes searched = b^d
- Alpha-Beta Search
 - Nodes searched $\sim 2b^{d/2}$ (optimal savings)
- Depth 7 search becomes depth 14
 - Bad program beats world champion

Recall MaxValue and MinValue

```
1. Function MaxValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or  $d > \text{maxD}$ )
3.     return eval(s)
4.    $v = -\text{infinity}$ 
5.   for (c in children(s))
6.      $v' = \text{MinValue}(c, \alpha, \beta, d+1)$ 
7.     if ( $v' > v$ )  $v = v'$ 
8.     if ( $v' \geq \beta$ ) return v
9.     if ( $v' > \alpha$ )  $\alpha = v'$ 
10.  return v
```

```
1. Function MinValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or  $d > \text{maxD}$ )
3.     return eval(s)
4.    $v = +\text{infinity}$ 
5.   for (c in children(s))
6.      $v' = \text{MaxValue}(c, \alpha, \beta, d+1)$ 
7.     if ( $v' < v$ )  $v = v'$ 
8.     if ( $v' \leq \alpha$ ) return v
9.     if ( $v' < \beta$ )  $\beta = v'$ 
10.  return v
```

eval(s) returns score w.r.t. the maximizing player

Alpha-Beta Pruning (minimax)

```
1.  Function AlphaBeta(s,  $\alpha$ ,  $\beta$ , d, max)
2.      if (terminal(s) or d > maxD)
3.          return eval(s)
4.      if (max) // maximizing player
5.          v = -infinity
6.          for (c in children(s))
7.              v' = MinValue(c,  $\alpha$ ,  $\beta$ , d+1)
8.              if (v' > v) v = v'
9.              if (v'  $\geq$   $\beta$ ) return v
10.             if (v' >  $\alpha$ )  $\alpha$  = v'
11.          return v
12.      else // minimizing player
13.          v = +infinity
14.          for (c in children(s))
15.              v' = MaxValue(c,  $\alpha$ ,  $\beta$ , d+1)
16.              if (v' < v) v = v'
17.              if (v'  $\leq$   $\alpha$ ) return v
18.              if (v' <  $\beta$ )  $\beta$  = v'
19.          return v
```

Can we shorten it?

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \max D$)
3. **return** eval(s)
4. $v = -\text{infinity}$
5. **for** (c in children(s))
6. $v' = \text{MinValue}(c, \alpha, \beta, d+1)$
7. **if** ($v' > v$) $v = v'$
8. **if** ($v' \geq \beta$) **return** v
9. **if** ($v' > \alpha$) $\alpha = v'$
10. **return** v

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
4. $v = -\text{infinity}$
5. **for** (c in children(s))
6. $v' = \text{MinValue}(c, \alpha, \beta, d+1)$
7. **if** ($v' > v$) $v = v'$
8. **if** ($v' > \alpha$) $\alpha = v'$
9. **if** ($v' \geq \beta$) **return** v
10. **return** v

Re-order the return statement, has same net effect

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
4. $v = -\text{infinity}$
5. **for** (c in children(s))
6. $v' = \text{MinValue}(c, \alpha, \beta, d+1)$
7. **if** ($v' > v$) $v = v'$
8. **if** ($v' > \alpha$) $\alpha = v'$
9. **if** ($\alpha \geq \beta$) **return** v
10. **return** v

Re-ordering allows us to compare α which will now store $\max(v')$

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
4. $v = -\text{infinity}$
5. **for** (c in children(s))
6. $v' = \text{MinValue}(c, \alpha, \beta, d+1)$
7. **if** ($v' > v$) $v = v'$
8. **if** ($v' > \alpha$) $\alpha = v'$
9. **if** ($\alpha \geq \beta$) **return** α
10. **return** v

We can also return α since it is storing $\max(v)$

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
4. v = -infinity
5. **for** (c in children(s))
6. v' = **MinValue**(c, α , β , d+1)
7. **if** (v' > v) v = v'
8. **if** (v' > α) α = v'
9. **if** ($\alpha \geq \beta$) **return** α
10. **return** v

v is now only used as a max calculation placeholder

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
- 4.
5. **for** (c in children(s))
6. $v' = \text{MinValue}(c, \alpha, \beta, d+1)$
- 7.
8. **if** ($v' > \alpha$) $\alpha = v'$
9. **if** ($\alpha \geq \beta$) **return** α
10. **return** α

We can completely remove v and use α

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
- 4.
5. **for** (c in children(s))
6. $\alpha = \max(\alpha, \text{MinValue}(c, \alpha, \beta, d+1))$
- 7.
- 8.
9. **if** ($\alpha \geq \beta$) **return** α
10. **return** α

we can now get rid of the placeholder variable v' as well

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
4. **for** (c in children(s))
5. $\alpha = \max(\alpha, \text{MinValue}(c, \alpha, \beta, d+1))$
6. **if** ($\alpha \geq \beta$) **return** α
7. **return** α

This is the most compact version of MaxValue

MaxValue

1. Function **MaxValue**(s, α , β , d)
2. **if** (terminal(s) or $d > \text{maxD}$)
3. **return** eval(s)
4. **for** (c in children(s))
5. $\alpha = \max(\alpha, \text{MinValue}(c, \alpha, \beta, d+1))$
6. **if** ($\alpha \geq \beta$) **break**
7. **return** α

We can use break instead of return, since we return outside the loop

MaxValue and MinValue

```
1. Function MaxValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   for (c in children(s))
5.      $\alpha = \max(\alpha, \text{MinValue}(c, \alpha, \beta, d+1))$ 
6.     if ( $\alpha \geq \beta$ ) break
7.   return  $\alpha$ 
```

```
1. Function MinValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   for (c in children(s))
5.      $\beta = \min(\beta, \text{MaxValue}(c, \alpha, \beta, d+1))$ 
6.     if ( $\beta \leq \alpha$ ) break
7.   return  $\beta$ 
```

Same shortening can be applied in MinValue algorithm

MaxValue and MinValue

```
1. Function MaxValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   for (c in children(s))
5.      $\alpha = \max(\alpha, \text{MinValue}(c, \alpha, \beta, d+1))$ 
6.     if ( $\alpha \geq \beta$ ) break
7.   return  $\alpha$ 
```

```
1. Function MinValue(s,  $\alpha$ ,  $\beta$ , d)
2.   if (terminal(s) or d > maxD)
3.     return eval(s)
4.   for (c in children(s))
5.      $\beta = \min(\beta, \text{MaxValue}(c, \alpha, \beta, d+1))$ 
6.     if ( $\alpha \geq \beta$ ) break
7.   return  $\beta$ 
```

Reverse comparison in MinValue to get the same one as MaxValue

Shortened Max/Min

```
1.  Function AlphaBeta(s,  $\alpha$ ,  $\beta$ , d, max)
2.      if (terminal(s) or d > maxD)
3.          return eval(s)
4.      if (max) // maximizing player
5.          for (c in children(s))
6.               $\alpha$  = max( $\alpha$ , AlphaBeta(c,  $\alpha$ ,  $\beta$ , d+1, !max))
7.              if ( $\alpha$   $\geq$   $\beta$ ) break
8.          return  $\alpha$ 
9.      else // minimizing player
10.         for (c in children(s))
11.              $\beta$  = min( $\beta$ , AlphaBeta(c,  $\alpha$ ,  $\beta$ , d+1, !max))
12.             if ( $\alpha$   $\geq$   $\beta$ ) break
13.         return  $\beta$ 
```

Notice Similarities

```
1.  Function AlphaBeta(s,  $\alpha$ ,  $\beta$ , d, max)
2.      if (terminal(s) or d > maxD)
3.          return eval(s)
4.      if (max) // maximizing player
5.          for (c in children(s))
6.               $\alpha$  = max( $\alpha$ , AlphaBeta(c,  $\alpha$ ,  $\beta$ , d+1, !max))
7.              if ( $\alpha \geq \beta$ ) break
8.          return  $\alpha$ 
9.      else // minimizing player
10.         for (c in children(s))
11.              $\beta$  = min( $\beta$ , AlphaBeta(c,  $\alpha$ ,  $\beta$ , d+1, !max))
12.             if ( $\alpha \geq \beta$ ) break
13.         return  $\beta$ 
```

There is duplicated code required due to if(max) being on the outside of the child loop

Bring Max Condition Inside the Loop

1. Function **AlphaBeta**(s, α , β , d, max)
2. **if** (terminal(s) or d > maxD)
3. **return** eval(s)
4. for (c in children(s))
5. **if** (max) $\alpha = \max(\alpha, \text{AlphaBeta}(c, \alpha, \beta, d+1, !\max))$
6. **if** (!max) $\beta = \min(\beta, \text{AlphaBeta}(c, \alpha, \beta, d+1, !\max))$
7. **if** ($\alpha \geq \beta$) **break**
8. **return** max ? α : β

Final shortened version of Alpha-Beta

Alpha-Beta (short version)

1. Function **AlphaBeta**(s, α , β , d, max)
2. **if** (terminal(s) or d > maxD)
3. **return** eval(s)
4. for (c in children(s))
5. **if** (max) $\alpha = \max(\alpha, \text{AlphaBeta}(c, \alpha, \beta, d+1, !\max))$
6. **if** (!max) $\beta = \min(\beta, \text{AlphaBeta}(c, \alpha, \beta, d+1, !\max))$
7. **if** ($\alpha \geq \beta$) **break**
8. **return** max ? α : β

Final shortened version of AB, returns VALUE of state

Result is exactly the same as the long version

Implement whichever one you feel more comfortable with

Which action do we take?

1. Function **AlphaBeta**(s, α , β , d, max)
2. **if** (terminal(s) or d > maxD)
3. **return** eval(s)
4. for (c in children(s))
5. **if** (max) $\alpha = \max(\alpha, \text{AlphaBeta}(c, \alpha, \beta, d+1, !\max))$
6. **if** (!max) $\beta = \min(\beta, \text{AlphaBeta}(c, \alpha, \beta, d+1, !\max))$
7. **if** ($\alpha \geq \beta$) **break**
8. **return** max ? α : β

So far only calculating value, need to record the action

Record Best Action

1. Function **AlphaBeta**(s, α , β , d, max)
2. **if** (terminal(s) or d > maxD)
3. **return** eval(s)
4. for (c in children(s))
5. $v' = \text{AlphaBeta}(c, \alpha, \beta, d+1, !\text{max})$
6. **if** (max) $\alpha = \max(\alpha, v')$
7. **if** (!max) $\beta = \min(\beta, v')$
8. **if** ($\alpha \geq \beta$) **break**
9. **return** max ? α : β

Re-introduce temporary value variable v' and is it for comparison

Record Best Action

1. Function **AlphaBeta**(s, α , β , d, max)
2. **if** (terminal(s) or d > maxD)
3. **return** eval(s)
4. for (c in children(s))
5. $v' = \text{AlphaBeta}(c, \alpha, \beta, d+1, !\text{max})$
6. **if** (max and ($v' > \alpha$)) $\alpha = v'$
7. **if** (!max and ($v' < \beta$)) $\beta = v'$
8. **if** ($\alpha \geq \beta$) **break**
9. **return** max ? α : β

Re-introduce temporary value variable v' and is it for comparison

Record Best Action

1. Function **AlphaBeta**(s, α , β , d, max)
2. **if** (terminal(s) or d > maxD)
3. **return** eval(s)
4. for (c in children(s))
5. $v' = \mathbf{AlphaBeta}(c, \alpha, \beta, d+1, !\max)$
6. **if** (max and ($v' > \alpha$))
7. $a = v'$
8. **if** (d == 0) bestAction = c.action
9. **if** (!max and ($v' < \beta$)) $\beta = v'$
10. **if** ($\alpha \geq \beta$) **break**
11. **return** max ? α : β

We can use a global/state variable to store best action at the root node (d=0)

Record Action AB Long Version

```
1.  Function AlphaBeta(s,  $\alpha$ ,  $\beta$ , d, max)
2.      if (terminal(s) or d > maxD)
3.          return eval(s)
4.      if (max) // maximizing player
5.          v = -infinity
6.          for (c in children(s))
7.              v' = AlphaBeta(c,  $\alpha$ ,  $\beta$ , d+1, !max)
8.              if (v' > v) v = v'
9.              if (v'  $\geq$   $\beta$ ) return v
10.             if (v' >  $\alpha$ )
11.                  $\alpha$  = v'
12.             if (d == 0) bestAction = c.action
13.          return v
14.      else // minimizing player
15.          v = +infinity
16.          for (c in children(s))
17.              v' = AlphaBeta(c,  $\alpha$ ,  $\beta$ , d+1, !max)
18.              if (v' < v) v = v'
19.              if (v'  $\leq$   $\alpha$ ) return v
20.              if (v' <  $\beta$ )  $\beta$  = v'
21.          return v
```

Time Limit

- What if we want a time limit?
- Search to a given depth time varies
- Incorporate iterative deepening
- Search $d=1, 2, \dots, \text{infinity}$
- When time runs out, return the best action from the last fully completed depth
- “Iterative Deepening Alpha Beta”

Iterative Deepening AlphaBeta

1. Function **IDAlphaBeta**(s, maxDepth)
2. depth = 0
3. bestAction = null
4. **while** (depth < maxDepth)
5. **try** { bestAction = AlphaBeta(s, -infinity, infinity, 0, true); }
6. **catch** (TimeoutException) { break; }
7. depth = depth + 1

AB With Time Limit

1. Function **AlphaBeta**(s, α , β , d, max)
2. **if** (terminal(s) or d > maxD) **return** eval(s)
3. **if** (timeElapsed > timeLimit) **throw** TimeOutException;
4. for (c in children(s))
5. $v' = \mathbf{AlphaBeta}(c, \alpha, \beta, d+1, !\text{max})$
6. **if** (max and ($v' > \alpha$))
7. $a = v'$
8. **if** (d == 0) bestAction = c.action
9. **if** (!max and ($v' < \beta$)) $\beta = v'$
10. **if** ($\alpha \geq \beta$) **break**
11. **return** max ? α : β

Throw timeout Exception to immediately halt AB execution