

# Branch and Bound

**ALGORITHMICS** 



Basic concepts



Examples of use

The problem of assigning tasks to agents

The problem of the puzzle

Optimal placement of rectangles

The problem of the n queens

BRANCH AND BOUND

### Basic concepts

### Branch and Bound

> This technique attempts to explore an implicit tree just like backtracking

Nodes represent the state of the problem

**Edges** represent valid changes

To solve the problem, we must find a node or a path in the associated graph

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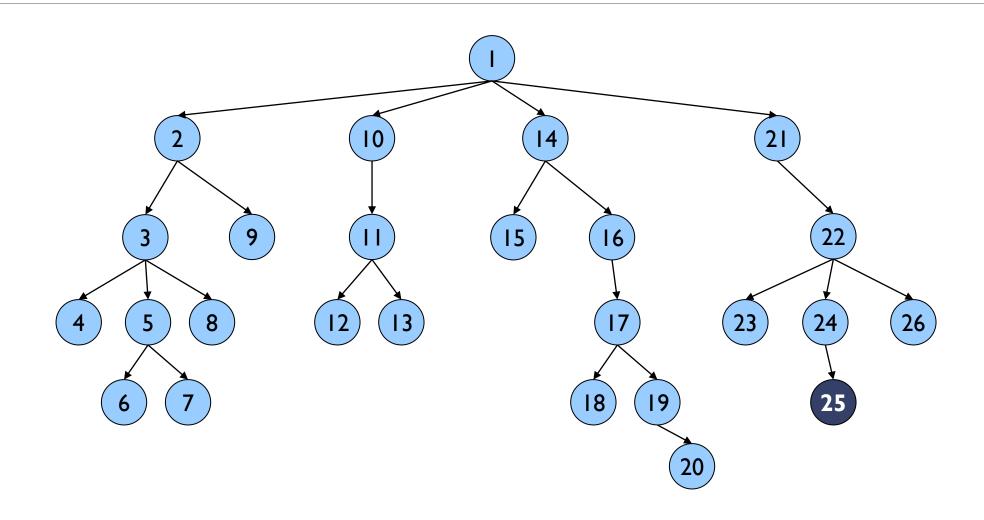
### Breadth-first search

With backtracking, the tree of states is developed in depth-first search

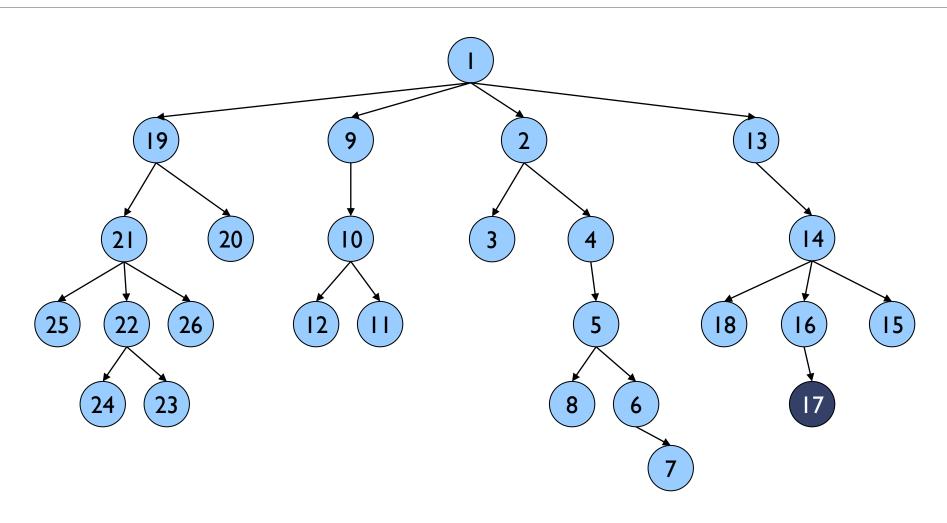
With branch and bound, all children nodes are generated before moving to the next state → It performs a breadth-first search

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### Depth-first search (backtracking)

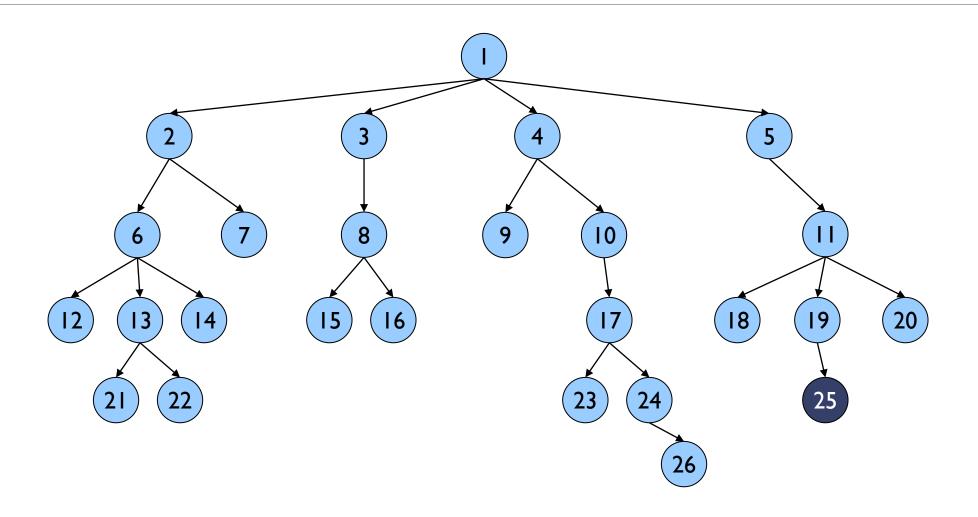


# Depth-first search (backtracking with some domain knowledge)

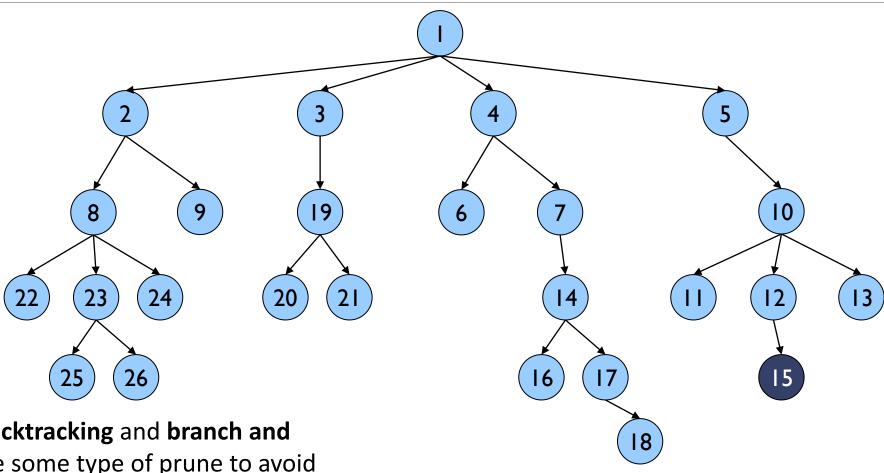


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### Breadth-first search (branch and bound)



# Breadth-first search (branch and bound with heuristic)



Both of them **backtracking** and **branch and bound** could use some type of prune to avoid developing bad known nodes

### Advantages of breadth-first search

> If we only search for the first solution, and there are solution states near the root

- With backtracking (depth-first) it is possible to enter a branch with many nodes that do not reach any solution or even that has no end (infinite)
  - This, with breadth-first, never happens

### **Branching Functions**

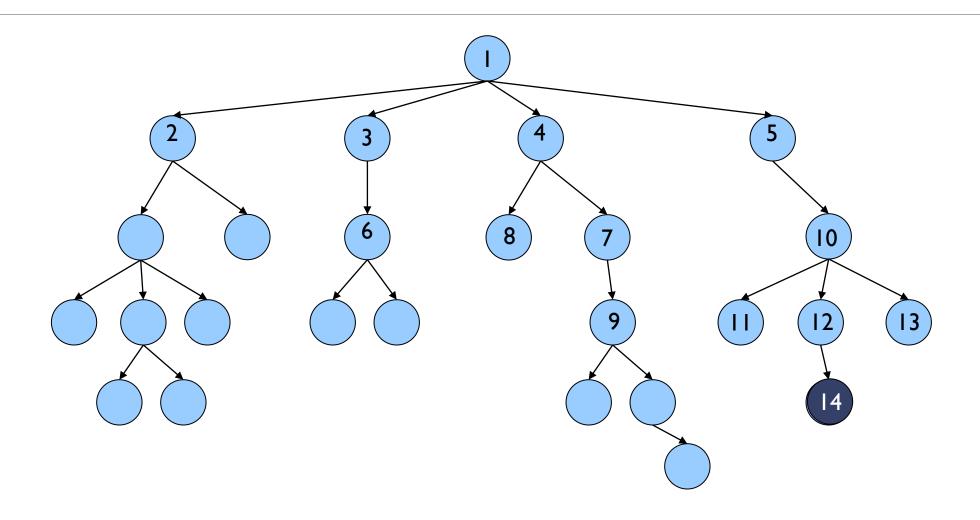
➤ The branching algorithms consist in developing nodes in the order indicated by a branching heuristic function

In every moment, we take as the node to be developed, the best that can be at any level of the tree

The branching algorithms make sense when the idea is to get the first solution and exit

> If you must develop the whole tree backtracking can be better

### Branching Functions (II)



### Branching data structure

The general scheme is the same as seen for exploration with breadth-first search, changing the FIFO queue by a **priority queue**, to store the "active" states to be developed

Priority queue (heap)

The heap, by structure (is an array), is limited to a maximum size that can only have up to a maximum pending to be developed live states

### Consequences of the fixed size of the heap

➤ We can only store a limited number of states

We are doing an uncontrolled pruning

- We could not store some states with a bad heuristic
- We may not find the solution even if it exists

### Pruning or bounding

We could add a **pruning heuristic function** to any of the already seen algorithms (depth-first, breadth-first, ...)

- The pruning function prevents the development of certain states that do nothing to find the solution to the problem:
  - Do not lead to any solution, or
  - Do not lead to better solutions, or
  - Are repeated developments

### Pruning heuristic

We assume that there is a boolean function that tells whether the children of a state can be pruned or not

```
if (node.getHeuristicValue() < pruneLimit)
  ds.insert(node);
else
  //do not store in the priority queue</pre>
```

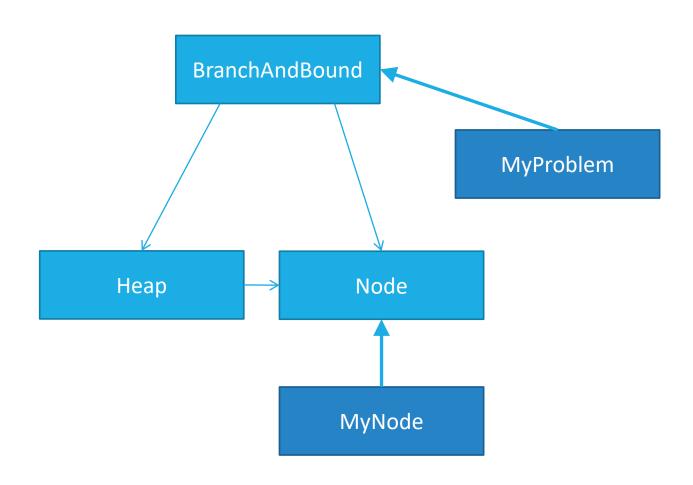
We can use this with depth-first or breadth-first algorithms

### General scheme

```
protected Heap ds; //nodes to be explored (not used nodes)
protected Node bestNode; //to save the final node of the best solution
protected Node rootNode; //initial node
public void branchAndBound(Node rootNode) {
     ds.insert(rootNode); //first node to be explored
     pruneLimit = rootNode.initialValuePruneLimit();
     while (!ds.empty() && ds.estimateBest() < pruneLimit) {</pre>
               Node node = ds.extractBestNode();
               ArrayList<Node> children = node.expand();
               for (Node child : children)
                         if (child.isSolution()) {
                                   int cost = child.getHeuristicValue();
                                   if (cost < pruneLimit) {</pre>
                                            pruneLimit = cost;
                                            bestNode = child;
                         else
                                  if (child.getHeuristicValue() < pruneLimit) {</pre>
                                            ds.insert(child);
      } //while
```

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### Relationship among classes



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BRANCH AND BOUND

Examples of use

## The problem of assigning tasks to agents

 We must assign j tasks to i agents, so that each agent performs only a task

 We have a matrix of costs, to know the cost of performing a task j by an agent i

 Goal: Minimizing the sum of the costs for executing the n tasks

	<u>j</u>	(tasks)	)	<b>→</b>	
		1	2		n
	а	11	12	•••	40
i (agents)	b	14	15	•••	22
ts) <b>*</b>	•••				
	n	17	14		28

21

### Branching heuristic



#### Calculation:

- The sum of the already assigned until that state, plus the best case (minimum of each column) of what remains to be assigned
- At each level we determine the assignation of another agent

#### Example:

• State a  $\rightarrow$  1, it carries a cost of **11**+14+13+22=60

#### Use of the heuristic:

 Explore the state of the tree that presents a lower heuristic value

	1		2	3	4
а	1	1	12	18	40
b	1	4	15	13	22
С	1	1	17	19	23
d	1	7	14	20	28

### Branching heuristic (II)



#### Calculation:

- The sum of the already assigned until that state, plus the best case (minimum of each column) of what remains to be assigned
- At each level we determine the assignation of another agent

#### **Example:**

• State  $a \rightarrow 2$  y  $b \rightarrow 1$ , it carries a cost of 12+14+19+23=68

#### Use of the heuristic:

 Explore the state of the tree that presents a lower heuristic value

	1	2	3	4
а	1	12	18	40
b	14	15	13	22
С	11	17	19	23
d	17	14	20	28

### Initial value prune limit



#### **Calculation:**

 Initially, the limit of pruning is the smallest of the sums of the two diagonals of the matrix of costs

#### Use:

 We prune every state that has a calculated value for the branching heuristic >= the limit of pruning

#### Change:

 The limit changes when we find a better solution

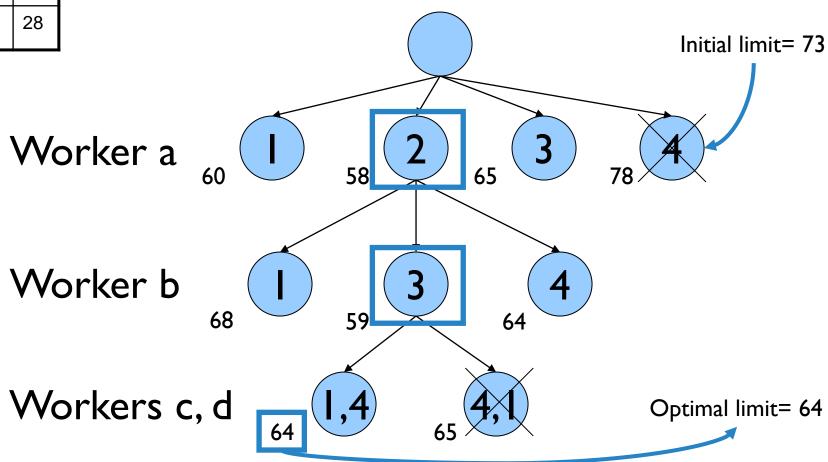
	1	2	3	4
а	11	12	18	40
b	14	15	13	22
С	11	17	19	23
d	17	14	20	28

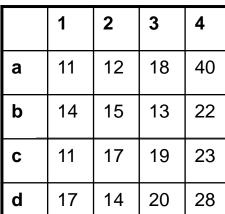
•Initial value limit: min(73,87)=73

	1	2	3	4
а	11	12	18	40
b	14	15	13	22
С	11	17	19	23
d	17	14	20	28

### Solution

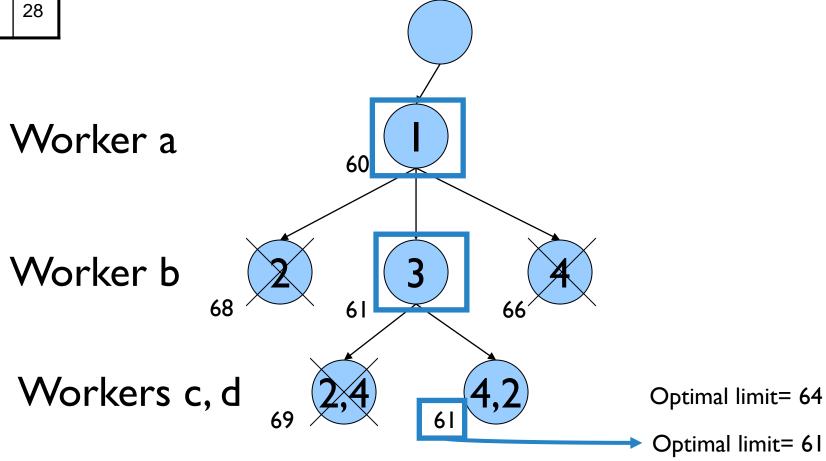






### Solution (II)

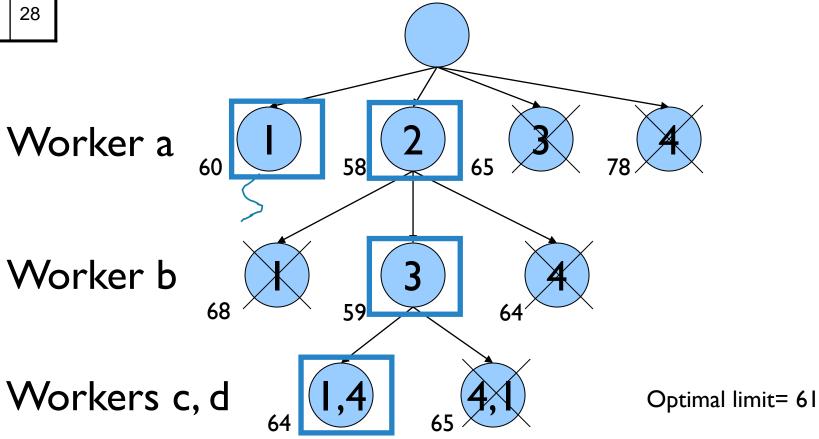




	1	2	3	4
а	11	12	18	40
b	14	15	13	22
С	11	17	19	23
d	17	14	20	28

### Solution (III)





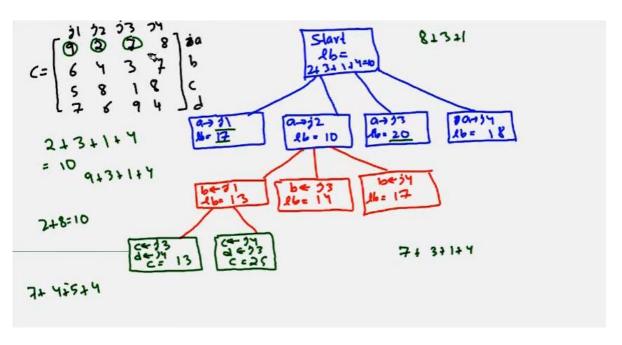
### Solution (IV)



- Optimal solution:
  - (a, 1), (b, 3), (c, 4), (d, 2)  $\rightarrow$  Total cost = 61

We only have developed 4 final states (6 if we count the two 2 diagonals to initialize the pruning threshold), instead of the 4!=24 that the backtracking development implies

> This algorithm is much more efficient in finding optimal solutions



### More at:

Assignment Problem using Branch and Bound <a href="https://www.youtube.com/watch?v=BV2MIZna6Pl">https://www.youtube.com/watch?v=BV2MIZna6Pl</a>



### The problem of the puzzle



This problem is developed on a board with 16 (n²) positions, where there are 15 (n²-1) pieces placed and there is an empty slot

Pieces cannot be lifted from the board, so their movement is only possible through slides on it

14	2	3	12
9		4	8
13	10	11	7
5	1	15	6



1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	

### Goal



The objective of the game is, from an initial state (disordered pieces), get a goal state, through a series of legal moves

Valid movements are those in which one piece, adjacent to a free position, is moved to that position

You can see these allowed movements as displacements from the empty cell to one of its four neighbors

### Strategy



- The full state space is composed of 16! states
  - It would be computationally unacceptable the cost to create and explore the tree

In fact, only half of these states are reachable from a given initial state, but still the number of states is still very large

We **must** be able to determine from a certain state, if it can reach a goal state or not

### Branching heuristic



- Example heuristics to develop states:
  - Smallest number of pieces placed on a wrong place
  - Smallest Manhattan distance of all the pieces regarding their final position
  - $\sum_{i=1}^{16} space(i)$ , where space(i) is the number of movements (distance) needed to put the piece i in the position i

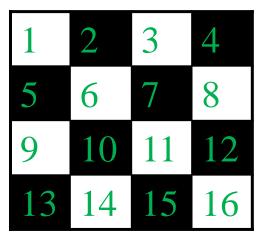
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### Pruning (bounding) heuristic



- Numbering boxes from I to I6, we define:
- position (i): position in the initial state of the piece number i
  - position (16) would be the position on the board of the empty piece
- smaller(i): is the number of pieces such that j<i and position(j)>position(i)

- The target state is reachable from a state if and only if
  - sum(smaller(i))+x is even
    - Where x is 1 If the empty cell is in one of the shaded cells and 0 if not



$$\sum_{i=1}^{16} smaller(i) + x$$

### Example



position(1)=1  position(2)=5  position(3)=2  position(4)=3
position(3)=2 position(4)=3
position(4)=3
. , ,
position(E)_7
position(5)=7
position(6)=10
position(7)=9
position(8)=13
position(9)=14
position(10)=15
position(11)=11
position(12)=8
position(13)=16
position(14)=12
position(15)=4
position(16)=6

smaller(1)=0	smaller(9)=0
smaller(2)=0	smaller(10)=0
smaller(3)=1	smaller(11)=3
smaller(4)=1	smaller(12)=6
smaller(5)=0	smaller(13)=0
smaller(6)=0	smaller(14)=4
smaller(7)=1	smaller(15)=11
smaller(8)=0	smaller(16)=10

$$\sum smaller(i) + x = 37 \quad (x = 0)$$

1	3	4	15
2		5	12
7	6	11	14
8	9	10	13

Therefore, the state **does not allow** to achieve the objective state

### Example (II)



position(1)=1 position(2)=2 position(3)=3 position(4)=4 position(5)=5 position(6)=6 position(7)=7 position(8)=8 position(9)=9 position(10)=10 position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15 position(16)=11	
position(3)=3  position(4)=4  position(5)=5  position(6)=6  position(7)=7  position(8)=8  position(9)=9  position(10)=10  position(11)=12  position(12)=16  position(13)=13  position(14)=14  position(15)=15	position(1)=1
position(4)=4 position(5)=5 position(6)=6 position(7)=7 position(8)=8 position(9)=9 position(10)=10 position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(2)=2
position(5)=5  position(6)=6  position(7)=7  position(8)=8  position(9)=9  position(10)=10  position(11)=12  position(12)=16  position(13)=13  position(14)=14  position(15)=15	position(3)=3
position(6)=6 position(7)=7 position(8)=8 position(9)=9 position(10)=10 position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(4)=4
position(7)=7 position(8)=8 position(9)=9 position(10)=10 position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(5)=5
position(8)=8 position(9)=9 position(10)=10 position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(6)=6
position(9)=9 position(10)=10 position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(7)=7
position(10)=10 position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(8)=8
position(11)=12 position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(9)=9
position(12)=16 position(13)=13 position(14)=14 position(15)=15	position(10)=10
position(13)=13 position(14)=14 position(15)=15	position(11)=12
position(14)=14 position(15)=15	position(12)=16
position(15)=15	position(13)=13
	position(14)=14
position(16)=11	position(15)=15
	position(16)=11

smaller(1)=0	smaller(9)=0	
smaller(2)=0	smaller(10)=0	
smaller(3)=0	smaller(11)=0	
smaller(4)=0	smaller(12)=0	
smaller(5)=0	smaller(13)=1	
smaller(6)=0	smaller(14)=1	
smaller(7)=0	smaller(15)=1	
smaller(8)=0	smaller(16)=5	

$$\sum smaller(i) + x = 8 \quad (x = 0)$$

1	2	3	4
5	6	7	8
9	10		11
13	14	15	12

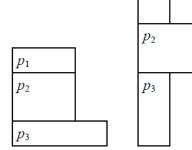
Therefore, the state **allows** to achieve the objective state

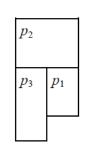
### Optimal placement of rectangles



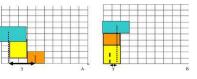


- Assuming we have n rectangular flat pieces  $p_1$ ,  $p_2$ , ...,  $p_n$ , each with an area  $(a_i, b_i)$  where  $(1 \le i \le n)$
- We want to fit them into a rectangular flat board
- > The problem consists in finding an arrangement of the n pieces so that the board we need to hold them has the minimum area
  - E.g., pieces are p1 = (1,2), p2 = (2,2) and p3 = (1,3)
  - The area of the covering boxes in each case is 12 (4x3), 14 (7x2), 10 (5x2), 9(3x3)





### Approach

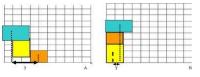


In each step k we need to explore a piece that has not yet been placed

For each of the pieces we have several options, since it can be placed around the pieces we have already placed both horizontally and vertically

The simplest way to represent the solution is a matrix of integer numbers, where 0 indicates an open position and a value k>0 indicates that this position is occupied by the piece  $p_k$ 

### Branching and bounding



- Every node will **generate** a **child** for each possible position of the next piece not yet included in the board
- > It is necessary to find the valid positions for the pieces
  - A valid position does not overlap with other pieces
  - A valid position is adjacent to other previously placed pieces
- We **prune** those nodes whose area so far exceeds that achieved by a previous solution
  - Remember we are looking for the minimum area
- Since we put a piece in each step, we have a **solution** when the level of the node is n
- The problem always has a solution because we assume that the board is large enough to accommodate all the pieces

### The problem of the n queens



The aim is to place n queens on a chessboard without any queen can capture another queen

> A queen can move any number of cells horizontally, vertically or diagonally

- Problem already solved with the backtracking technique
  - Now we will generate only those states in which queens do not eat other queens

### Branching heuristic



- Develop the state for which we will have more queens placed
- In case of a tie, we put a queen on a cell whose longest diagonal length is the lowest one
  - diagonal(i, j), length of the longest diagonal passing through the cell(i, j)
  - The table at right shows the values of the function in each cell

5	4	3	4	5
4	5	4	5	4
3	4	5	4	3
4	5	4	5	4
5	4	3	4	5

#### Assigning numerical values



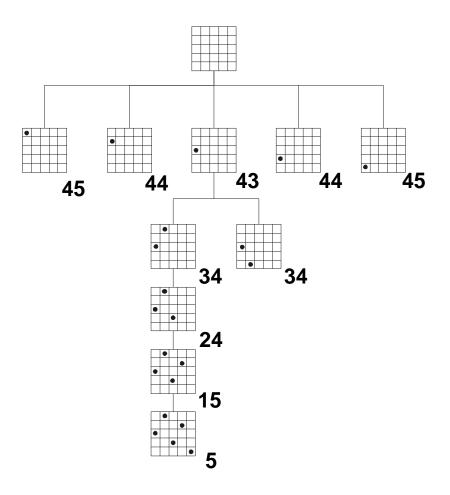
- ➤ We seek a function → the lower the value, the better the state to be developed
  - State with more placed queens:
    - $n n^{o}$  placed queens
  - In case of a tie, queen in a diagonal cell *diagonal(i,j)* which value is the lowest

- To merge the terms:
  - $(n n^{o} placed queens) * 10 + diagonal(i,j)$

#### General idea



The "living" states are the nodes generated by the heuristic described and stored in the priority queue





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### If we only search for the first solution, and there are solution states near the root, the best choice is usually Backtracking

1. True





With Backtracking (depth-first) it is possible to enter a branch with many nodes that do not reach any solution or even that has no end (infinite)



1 True

2 False



## If you must develop the whole tree, Backtracking is usually better than Branch and bound



1. True

2. False

## Which data structure would be most appropriate for Branch and bound (without branching function) using Java as the programming language?

- 2. PriorityQueue
- 3. List
- 4. TreeSet
- 5. HashMap
- 6. ArrayList
- √ 7. Queue



### Working with Backtracking, the tree of states is developed in



1. Depth-first search

2. Breadth-first search

#### POLL OPEN

## With which technique we could use some kind of pruning to avoid developing bad known nodes?

- 1. Backtracking
- 2. Branch and bound



3. Both of them



### A branching algorithm makes more sense when the idea is to get the first solution and exit

**V** 1

1. True

2. False



# The pruning function prevents the development of certain states that do nothing to find the solution to the problem when they:

- 1. Are repeated developments
- 2. Do not lead to better solutions
- 3. Do not lead to any solution
- √ 4. All of them



### Working with Branch and bound, the tree of states is developed in

1. Depth-first search



✓ 2. Breadth-first search



# Which data structure would be most appropriate for Branch and bound (with branching function) using Java as the programming language?

2 Queue



3 PriorityQueue

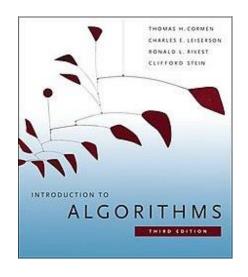
4 ArrayList

5 TreeSet

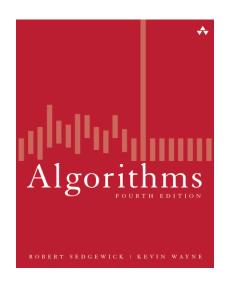
6 Stack

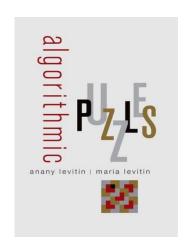
7 HashMap

### Bibliography

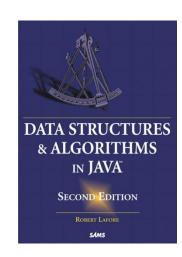


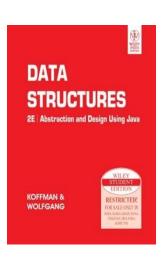












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