### 13 - Backtracking

[KOMS119602] & [KOMS120403]

Design and Analysis of Algorithm (2021/2022)

Dewi Sintiari

Prodi S1 Ilmu Komputer Universitas Pendidikan Ganesha

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### Principal of Backtracking

- The exhaustive-search technique suggests generating all candidate solutions and then identifying the one (or the ones) with a desired property.
- Backtracking algorithm improves exhaustive search.
- In exhaustive search, all possible solutions are explored and evaluated one-by-one
- In backtracking, we do not examine all possibilities, only the possibilities that lead to the solution. Other nodes that do not lead to the solution are pruned.

#### Central idea:

To cut off a branch of the problems state-space tree, as soon as we can deduce that it cannot lead to a solution.



# Principal of Backtracking

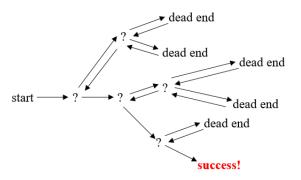


Figure: Illustration of backtracking (sumber: https://miro.medium.com/)

### Representation of solution

- Representation: an output can be thought of as n-tuple  $(x_1, x_2, \ldots, x_n)$  where each coordinate  $x_i$  is an element of some finite linearly ordered set  $S_i$ .
- Tuples: all solution tuples can be of the same length (the n-queens and the Hamiltonian circuit problem) and of different lengths (the Subset-sum problem).

### Backtracking in DFS

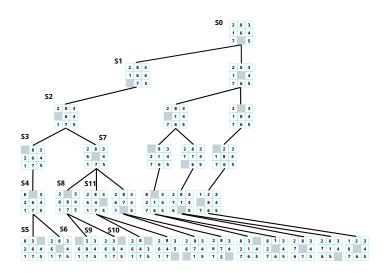
Backtracking in DFS is used in solution-searching problems that have many possibilities of solution.

The solution is obtained by looking in a depth-first approach

- You do not have enough information to know the next step.
- Each decision leads you to several/many new choices.
- Several sequence of choices may be the problem's solution.

In DFS, backtracking is used as a methodological way to try several sequences of decision.

### Example of backtracking in DFS



# State-space tree

### State-space tree (1)

Backtracking can be seen as searching in a tree from the root to the leaves (solution node).

### State-space tree

It is a tree representing all the possible states (solution or non-solution) of the problem from the root as an initial state to the leaf as a terminal state.

A backtracking algorithm generates, explicitly or implicitly, a **state-space tree**:

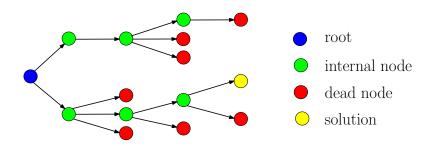
- its nodes represent partially constructed tuples with the first i coordinates defined by the earlier actions of the algorithm;
- if such a tuple  $(x_1, x_2, \ldots, x_i)$  is not a solution, the algorithm finds the next element in  $S_{i+1}$  that is consistent with the values of  $(x_1, x_2, \ldots, x_i)$  and the problems constraints, and adds it to the tuple as its (i+1)st coordinate;
- if such an element does not exist, the algorithm backtracks to consider the next value of  $x_i$ , and so on.



# State-space tree (2)

- Root represents an initial state before the search begins;
- Internal nodes
  - the nodes of the first level in the tree represent the choices made for the first component of a solution;
  - the nodes of the second level represent the choices for the second component;
  - and so on...;
- Leaves represent either non-promising dead ends or complete solutions found by the algorithm.

# State-space tree (3)



### Types of nodes in the state-space tree

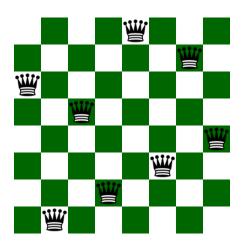
- Promising node: corresponds to a partially constructed solution that may still lead to a complete solution;
- Non-promising node: dead node

### State-space tree (4)

- The solution is searched by generating the state nodes, so that it produces paths from the root to the leaves;
- To generate the nodes, the DFS rule is followed;
- The generated nodes are called live node;
- The live node being expanded is called expand-node;
- Each time the expand-node is expanded, the generated path gets longer;
- Function that is used to "kill" an expand-node is called bounding function;
- When a node is killed, then automatically all its children nodes are pruned;
- If the paths-generation ends up with dead node, the searching is backtrack to the parents nodes;
- These parents nodes become the new expand-nodes;
- The searching is stopped if we find a solution.



# *n*-Queens problem

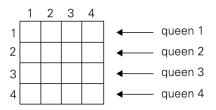


### *n*-Queens problem

#### Problem

Place n queens on an  $n \times n$  chessboard, so that no two queens attack each other (by being in the same row, in the same column, or in the same diagonal).

- For n = 1, the problem has a trivial solution
- For n = 2, 3, the problem has no solution
- What if n = 4?



### Algorithm

#### **START**

- 1 begin from the leftmost column
- ② if all the queens are placed, return true/ print configuration
- 3 check for all rows in the current column
  - if queen placed safely, mark row and column; and recursively check if we approach in the current configuration, do we obtain a solution or not
  - if placing yields a solution, return true
  - if placing does not yield a solution, unmark and try other rows
- if all rows tried and solution not obtained, return false and backtrack

#### **END**



### *n*-Queens problem

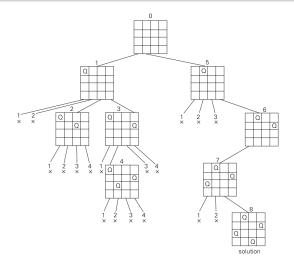


Figure: State-space tree of solving the four-queens problem by backtracking;  $\times$  denotes an unsuccessful attempt to place a queen in the indicated column. The numbers above the nodes indicate the order in which the nodes are generated.

### *n*-Queens problem

- We start with the empty board and then place queen 1 in the first possible position of its row, which is in column 1 of row 1.
- ② Then we place queen 2, after trying unsuccessfully columns 1 and 2, in the first acceptable position for it, which is square (2,3), the square in row 2 and column 3.
- This proves to be a dead end because there is no acceptable position for queen 3.
- So, the algorithm backtracks and puts queen 2 in the next possible position at (2,4).
- **5** Then queen 3 is placed at (3,2), which proves to be another dead end.
- **1** The algorithm then backtracks all the way to queen 1 and moves it to (1,2).
- Queen 2 then goes to (2,4), queen 3 to (3,1), and queen 4 to (4,3), which is a solution to the problem.

# Other problems

### Hamiltonian circuit problem

#### Problem

Given a connected graph G, find a Hamiltonian circuit in G. (Recall that a Hamiltonian circuit is a circuit that visits all vertices of G exactly once.)

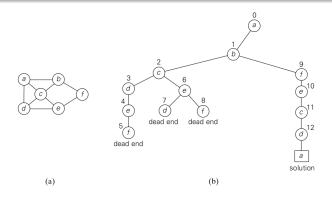


Figure: (a) Graph. (b) State-space tree for finding a Hamiltonian circuit. The numbers above the nodes of the tree indicate the order in which the nodes are generated.

# Subset-sum problem (1)

#### Problem

Find a subset of a given set  $A = \{a_1, ..., a_n\}$  of n positive integers whose sum is equal to a given positive integer d.

**Example 1**: Given  $A = \{1, 2, 5, 6, 8\}$ , d = 9, the solution are:  $\{1, 2, 6\}$  and  $\{1, 8\}$ .

**Example 2:** Given  $A = \{3, 5, 6, 7\}$ , d = 15, the solution are:  $\{3, 5, 7\}$ .

# Subset-sum problem (2)

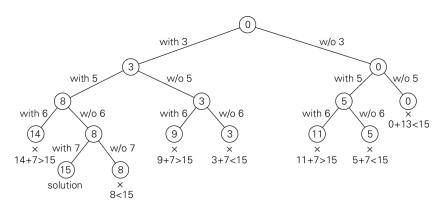


Figure: Complete state-space tree of the backtracking algorithm applied to the instance  $A = \{3, 5, 6, 7\}$  and d = 15 of the Subset-sum problem. The number inside a node is the sum of the elements already included in the subsets represented by the node. The inequality below a leaf indicates the reason for its termination.

### Subset-sum problem (1)

A path from the root to a node on the *i*th level of the tree indicates which of the first *i* numbers have been included in the subsets represented by that node.

We record the value of s, the sum of these numbers, in the node.

- If s is equal to d, we have a solution to the problem. We can either report this result and stop or,
- If all solutions need to be found, continue by backtracking to the nodes parent.
- If s is not equal to d, we can terminate the node as non-promising if either of the following two inequalities holds:

$$s + a_{i+1} > d$$
 (the sum  $s$  is too large)  $s + \sum_{j=i+1}^n a_j < d$  (the sum  $s$  is too small)

# Backtracking framework

### Backtracking algorithm framework

### Algorithm 1 Backtracking

```
1: procedure Backtrack(X[1..i])
       input: X[1..i]: the first i promising components of a solution
 2:
 3:
       output: all the tuples representing the problem's solution
 4:
       if X[1..i] is a solution then
 5:
           write (X[1..i])
 6:
       else
 7:
           for each x \in S_{i+1} consistent with X[1..i] and the constraints do
 8.
               X[i+1] \leftarrow x
 9:
               Backtrack(X[1..i+1])
10:
           end for
11:
        end if
12: end procedure
```

### Time complexity

Backtracking is basically an exhaustive search performed over the search space. So the time complexity of a backtracking algorithm is **defined by the size of the search space**.

For example, in the *n*-queens problem and Hamiltonian problem, the size of the search space is about  $\mathcal{O}(n!)$ .

Intuitively, the first queen has n placements, the second queen must not be in the same column as the first as well as at an oblique angle, so the second queen has n-1 possibilities, and so on, with a time complexity of  $\mathcal{O}(n!)$ .

### Advantages & drawbacks

#### **Advantages**

- Typically applied to difficult combinatorial problems for which no efficient algorithms for finding exact solutions possibly exist.
- Unlike the exhaustive-search approach, backtracking at least holds a
  hope for solving some instances of nontrivial sizes in an acceptable
  amount of time (especially for optimization problems).
- Even if backtracking does not eliminate any elements of a problems state space and ends up generating all its elements, it provides a specific technique for doing so.

#### **Drawbacks**

- Backtracking is **not** a very efficient technique (even though it was succeeded to use in the previous problems).
- In the worst case, it may have to generate all possible candidates in an exponentially (or faster) growing state space of the problem.