Design and Analysis of Algorithm Backtrack (I)

- Classical Examples
- Principles of Backtrack
- 3 Loading Problem
- 4 Graph Coloring Problem
- Estimation of Leaves

Backtrack Paradigm

Recursive approach is essentially traveling the whole tree defined by the recusive relation.

 The subtrees may repeat, so we need to cache intermediate results to improve efficiency. This is exactly the essence of dynamic programming.

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 The subtrees may repeat, so we need to cache intermediate results to improve efficiency. This is exactly the essence of dynamic programming.

For some problems, the subtrees will not overlap.

- In such case, there is no better algorithm other than traveling the entire tree. But, we can travel the entire tree smartly.
- This is what backtrack technique concerns: stop visiting the subtree if the solution won't appear and backtrack to the parent node
 - basic backtrack strategy: Domino property defined by problem constraint
 - advanced backtrack strategy: branch-and-bound



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Example 1: Eight Queen Problems

Eight queens puzzle. Placing eight chess queens on an 8×8 chessboard so that no two queens threaten each other.

 a solution requires that no two queens share the same row, column, or diagonal.

Eight queens puzzle is a special case of the more general n queens problem: placing n non-attacking queens on an $n \times n$ chessboard.



Counting Solutions

Solution is an n-dimension vector over [n]: exist for all natural numbers n with the exception of n=2,3.

- Eight queens puzzle has 92 distinct solutions, the entire solution space is $C_{64}^8 = 4,426,165,368$.
- If solutions that differ only by the <u>symmetry operations of</u> <u>rotation and reflection of the board are counted as one, the</u> <u>puzzle has 12 solutions, called as fundamental solutions.</u>

n	fundamental	all
8	12	92
9	46	352
10	92	724
26	2,789,712,466,510,289	22,317,699,616,364,044
27	29, 363, 495, 934, 315, 694	234,907,967,154,122,528

Background of Eight Queen Puzzle

Origin of Eight Queen Puzzle

Max Bezzel first proposed this problem in 1848, Frank Nauck gave the first solution in 1850 and extended it n queen puzzle. Many mathematicians including Carl Guass also studied this problem.

Edsger Dijkstra exemplified the power of *depth-first backtracking algorithm* via this problem.

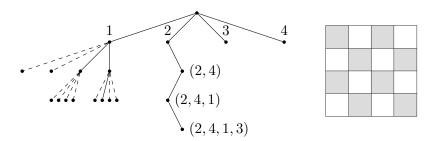
There is no known formula for the exact number of solutions, or even for its asymptotic behavior. The 27×27 board is the highest-order board that has been completely enumerated.

How to solve?

- modeling all possible solutions as leaf nodes of a tree
- traversal the solution space via traveling the tree



Demo of Quadtree for 4 Queens Puzzle



Traversal the tree via depth-first order to find all solutions

- *i*-th level node represent *i*-th element in solution vector
- \bullet each node has 4 children, representing possible choices from 1,2,3,4 in the next column
- leaves correspond to solutions
- ullet in the i-th level, the branching choice is less than n-i



Example 2: 0-1 **Knapsack Problem**

Problem. Given n items with value v_i and weight w_i , as well as a knapsack with weight capacity W. The number of each item is 1. Find a solution that maximize the value.

Solution. n dimension vector $(x_1,x_2,\ldots,x_n)\in\mathbb{Z}_2^n$, $x_i=1\Leftrightarrow$ selecting item i

Nodes: (x_1, x_2, \dots, x_k) corresponds to partial solution

Search space. In all level, the branching choice is always $2 \rightsquigarrow$ perfect binary tree with 2^n leaves

Candidate solution. Satisfy constraint $\sum_{i=1}^{n} w_i x_i \leq W$

Optimal solution. The candidate solutions that achieve maximal values.



A Demo

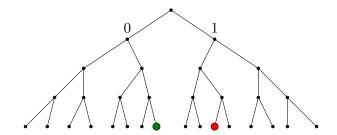
Table: n = 4, W = 13

item	1	2	3	4
value	12	11	9	8
weight	8	6	4	3

Two candidate solutions

- (0,1,1,1): v=28, w=13
- (1,0,1,0): v=21, w=12

Optimal solution is (0, 1, 1, 1)



Example 3: Traversal Salesman Problem

Problem. Given n cities $C = \{c_1, c_2, \dots, c_n\}$ and $d(c_i, c_j) \in \mathbb{Z}^+$. Find a cycle with minimal length that travels each city once.

Solution. A permutation of (1, 2, ..., n) — $(k_1, k_2, ..., k_n)$ such that

$$\min \left\{ \sum_{i=1}^{n-1} d(c_{k_i}, c_{k_{i+1}}) + d(c_{k_n}, c_{k_1}) \right\}$$

$$C = \{1, 2, 3, 4\}$$

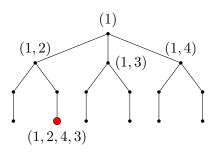
$$d(1, 2) = 5, d(1, 3) = 9$$

$$d(1, 4) = 4, d(2, 3) = 13$$

$$d(2, 4) = 2, d(3, 4) = 7$$

Solution is (1, 2, 4, 3), length of cycle is 5 + 2 + 7 + 9 = 23

Search Space of TSP



Any node can searve as the root, cause TSP is defined over a undirected graph.

Search space. In the i-th level, the branching choice is always n-i \leadsto perfect binary tree with (n-1)! leaves \leadsto number of all possible permutations over $\{1,\dots,n\}$

Summary

Classical examples of Backtrack

• n queens puzzle, 0-1 knapsack, TSP

Solution: vector

Search space: tree

 nodes correspond to partial solutions, leaves correspond to candidate solutions

Search order: depth-first, breadth-first, jump-hop

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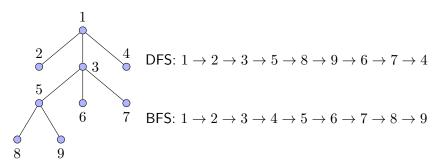
Main Idea of Backtrack

Scope of application. Search or optimization problem

Search space. Tree

- leaves: candidate solution
- nodes: partial solution

How to search. Systematically traversal the tree: DFS, BFS, ...



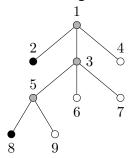
States of Nodes

The tree is explored dynamically. Let v be the candidate node (corresponding to partial solution) and P be the predicate that checks if v satisfies constraint.

- $P(v) = 1 \Rightarrow \mathsf{expand}$
- $P(v) = 0 \Rightarrow \text{backtrack to parent node}$

States of node

- white: unexplored
- gray: visiting its subtree
- black: finishing the traversal of this subtree



DFS: $1 \rightarrow 2 \rightarrow 3 \rightarrow 5 \rightarrow 8$

- finished visiting: 2,8
- being visited: 1, 3, 5
- unexplored: 9, 6, 7, 4



Basic Backtrack Technique: Domino Property

At node $v = (x_1, \ldots, x_k)$

$$P(x_1,\ldots,x_k)=1\Leftrightarrow (x_1,\ldots,x_k)$$
 meet some property

Example. n queens puzzle, placing k queens in positions without attacking each other

Domino property → admit safe backtrack

$$P(x_1, x_2, \dots, x_{k+1}) = 1 \Rightarrow P(x_1, x_2, \dots, x_k) = 1, 0 < k < n$$

Converse-negative proposition

$$P(x_1, x_2, \dots, x_k) = 0 \Rightarrow P(x_1, x_2, \dots, x_{k+1}) = 0, 0 < k < n$$

k-dimension vector does not satisfy constraint \Rightarrow its k+1-dimension extension does not satisfy constraint either

- this property guarantees that backtracking will not miss any solution
- safely backtrack when $P(x_1, x_2, \dots, x_k) = 0$



A Counterexample

Find integer solutions for inequality

$$5x_1 + 4x_2 - x_3 \le 10, 1 \le x_k \le 3, k = 1, 2, 3$$

$$P(x_1, \dots, x_k) = 1 \text{ iff } \sum_{i=1}^k a_i x_i \le 10$$

Does not satisfy Domino property

$$5x_1 + 4x_2 - x_3 \le 10 \Rightarrow 5x_1 + 4x_2 \le 10$$

A Counterexample

Find integer solutions for inequality

$$5x_1 + 4x_2 - x_3 \le 10, 1 \le x_k \le 3, k = 1, 2, 3$$

$$P(x_1, ..., x_k) = 1 \text{ iff } \sum_{i=1}^k a_i x_i \le 10$$

Does not satisfy Domino property

$$5x_1 + 4x_2 - x_3 \le 10 \Rightarrow 5x_1 + 4x_2 \le 10$$

Modification to satisfy Domino property: set $x_3^\prime=3-x_3$

$$5x_1 + 4x_2 + x_3' \le 13, 1 \le x_1, x_2 \le 3, 0 \le x_3' \le 2$$

Summary

The premise condition to use backtrack: Domino property

General steps of backtrack algorithm

- Define solution vector (include the range of every element), $(x_1, x_2, \ldots, x_n) \in X_1 \times \cdots \times X_n$
- After fixing $(x_1, x_2, \dots, x_{k-1})$, update admissible range of x_k as $A_k \subseteq X_k$
- Decide if Domino property is satisfied
- Decide the search strategy: DFS, BFS
- Decide the data structure to store the search path



Backtrack Recursive Template

Algorithm 1: BackTrack(n) //output all solutions

- 1: **for** k = 1 to n **do** $A_k \leftarrow X_k$; //initialize
- 2: ReBack(1);

Algorithm 2: ReBack(k) //k is the current depth of recursion

```
1: if k=n then return solution (x_1,\ldots,x_n);
2: else
3: while A_k \neq \emptyset do
4: x_k \leftarrow A_k //according to some order;
```

- 5: $A_k \leftarrow A_k \{x_k\};$
- 6: update A_{k+1} , ReBack(k+1);
- 7: **end**
 - The above is the oversimplify pseudocode.
 - One must be careful when dealing with domains A_k and solution vector x when coding (value transfer vs. reference transfer)

Backtrack Iterative Template

Algorithm 3: BackTrack(n) //all solutions

```
1: for k=1 to n do A_k \leftarrow X_k; //initialize

2: k \leftarrow 1;

3: while A_k \neq \emptyset do

4: x_k \leftarrow A_k; A_k \leftarrow A_k - \{x\};

5: if k < n then k \leftarrow k + 1;

6: else (x_1, x_2, \dots, x_n) is solution;

7: end

8: if k > 1 then k \leftarrow k - 1; goto 3;
```

- A_k is determined by (x_1, \ldots, x_{k-1})
- The algorithm terminates when all A_i are empty. Otherwise, it will backtrack (line 8).

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Loading Problem

Problem. Given n containers with weight w_i , two boats with weight capacity W_1 and W_2 s.t. $w_1 + \cdots + w_n \leq W_1 + W_2$.

Goal. If there exists a scheme to load the n containers on two boats. Please give a scheme if it is solvable.

Example

• $w_1 = 90, w_2 = 80, w_3 = 40, w_4 = 30, w_5 = 20, w_6 = 12, w_7 = 10, W_1 = 152, W_2 = 130$

Solution: load 1, 3, 6, 7 on boat 1 and the rest on boat 2

Main idea: Let the total weights be W.

- Load on boat 1 first. Using backtrack to find a solution that maximizes W_1^* , where W_1^* is the real capacity.
- ② Then check if $W-W_1^* \leq W_2$. Return "yes" if true and "no" otherwise.

Pseudocode

Algorithm 4: Loading (W_1)

```
1: W_1^* \leftarrow 0; C \leftarrow 0; i \leftarrow 1;

2: while i \leq n do //line 3-4: whether to load container i

3: if C + w_i \leq W_1 then C \leftarrow C + w_i, x[i] \leftarrow 1, i = i + 1;

4: else x[i] \leftarrow 0, i \leftarrow i + 1;

5: end

6: if W_1^* < C then record solution, W_1^* \leftarrow C;

7: while i > 1 and x[i] = 0 do i - 1; //find a backtrack node

8: if i = 0 then return optimal solution; //backtrack to root

9: else x[i] = 0; C \leftarrow C - w_i; i = i + 1, goto 2; //x[i] = 1:
```

line 7-9: find a backtrack point

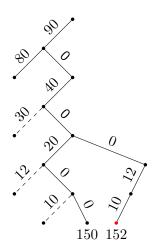
- back to the root: travel all the tree (no left branch)
- 2 find a left branch: change it to right branch

$$w_1 = 90, w_2 = 80, w_3 = 40, w_4 = 30, w_5 = 20, w_6 = 12, w_7 = 10$$

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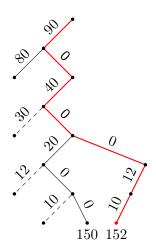
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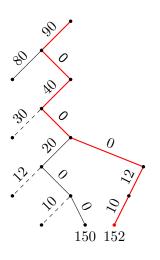
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it is loadable

- 1, 3, 6, 7 on boat 1
- 2, 4, 5 on boat 2

time complexity $W(n) = O(2^n)$

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Graph Coloring Problem

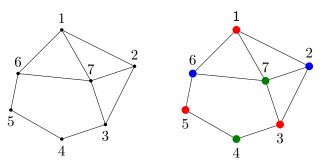
Problem. Undirected graph G and m colors. Coloring the vertices to ensure the connected two vertices with different color.

Goal. Output all possible coloring schemes. Output "no" if there is none.

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$$n = 7, m = 3$$

Algorithm Design

Input. G=(V,E), $V=\{1,2,\ldots,n\}$, color set $\{1,2,\ldots,m\}$ Solution vector. (x_1,x_2,\ldots,x_n) , $x_i\in[m]$ (x_1,\ldots,x_k) gives partial solution for vertice set $\{1,2,\ldots,k\}$ Search tree. m-fork tree

Scarch tree. W Tork tree

Constraint. At node (x_1, \ldots, x_k) , the colors of vertices in adjacent list of node k+1 are not usable.

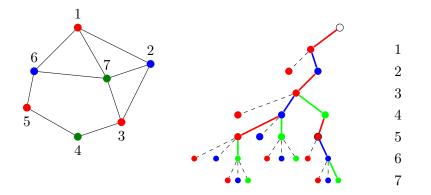
• If the nodes in adjacent list have used up m colors, then node k+1 is not colorable. In this case, back to parent node. (Domino property obviously holds)

Search strategy: DFS

Time complexity: $O(nm^n)$

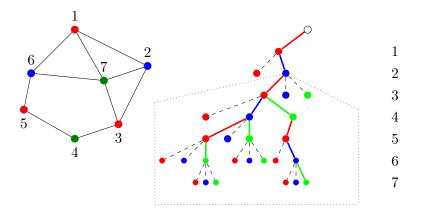
- the depth of tree is $n \Rightarrow$ totally at most m^n leaves
- every step need to find usable colors \Rightarrow require O(n) cost





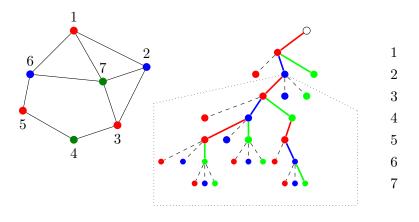
the first solution vector: (1,2,1,3,1,2,3)

The Structure of Search Tree



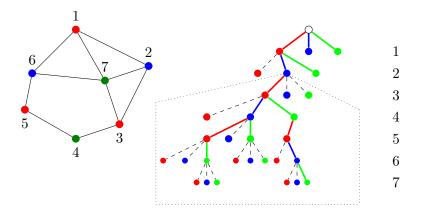
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The Structure of Search Tree



The first solution vector: (1, 2, 1, 3, 1, 2, 3)

Time complexity

Time complexity: $O(nm^n)$

Symmetry \sim only need to search at most 1/6 solution space

- the permutation over (1,2,3) is $6 \sim$ for any specific solution, there exist 6 homogeneous solution
- level-2 has 2-fold solution (e.g. color blue and green are exchangeable), level-1 has 3-fold solution (node 1 can pick red, green or blue); the closer to the root, the more flexibility of enchantment.

Additional reasoning also helps to reduce search scope

- Example: if node 1,2,3 have been colored differently, then node 7 is definitely non-colorable because it connects with node $1,2,3 \leadsto$ backtrack from this node
- Need trade-off between search and decide



Applications of Graph Coloring

Arrangement of meeting room

There are n events to be arranged, if the slots of event i and event j overlap, we say i and j are not compatible. How to arrange these events with smallest number of meeting rooms?

Modeling

- ullet Treat event as node, if i,j are not compatible, then add an edge between i and j.
- Treat meeting rooms as colors.

The arrangement problem is transformed to finding a coloring scheme with smallest colors.

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Estimation of Leaves

Sometimes, we need to know the size of problems (captured by the number of leaves)

Finding the exact number may require travel the whole tree exhaustively, which is equivalent to solve the problem.

Monte Carlo method

- ① Choose a random path from root until there is no more branching, i.e., randomly and sequentially assign values to x_1, x_2, \ldots , until the vector cannot be further expanded.
- ② Assume other $|A_i|-1$ branches has the same path as selected one, calculate the nodes of search tree
- 3 Repeat step 1 and 2, compute the average number of nodes.

Estimate *n* **Queen Puzzle**

4: $\ell \leftarrow \ell + m$:

5: **end** 6: $\ell \leftarrow \ell/t$;

```
Algorithm 5: MonteCarlo(n,t)

Input: n=\# number of queens, t=\# number of sampling

Output: \ell, average length of t times sampling

1: \ell \leftarrow 0;

2: for i=1 to t do //sampling time is t

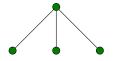
3: m \leftarrow \text{Estimate}(n); //number of nodes;
```

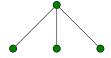
One Sampling

Parameter

- ullet is the number of nodes of current sampling
- k is the depth
- r_{prev} : # (nodes on the previous level)
- r_{current} : # (nodes on the current level)
- $r_{\mathsf{current}} = r_{\mathsf{prev}} \times \#(\mathsf{branches})$
- n is the depth of tree

Computation oder: randomly select until reaching the leaves





$$r_{\text{prev}} = 2$$
, $r_{\text{current}} = r_{\text{prev}} \cdot 3 = 6$

Pseudocode

Algorithm 6: Estimate(n)

```
1: \ell \leftarrow 1; r_{\mathsf{prev}} \leftarrow 1; k \leftarrow 1; //\ell initially is the root node;

2: while k \leq n do

3: if A_k = \emptyset then return \ell; //no more branch

4: x_k \overset{\mathbb{R}}{\leftarrow} A_k //randomly select a branch;

5: r_{\mathsf{current}} \leftarrow r_{\mathsf{prev}} \times |A_k|;

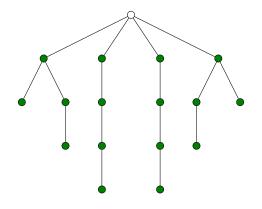
6: \ell \leftarrow \ell + r_{\mathsf{current}};

7: r_{\mathsf{prev}} \leftarrow r_{\mathsf{current}};

8: k \leftarrow k + 1;

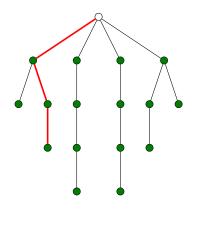
9: end
```

Real Case: 4-Queens Puzzle



17 nodes

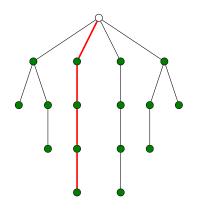
Random Selected Path 1



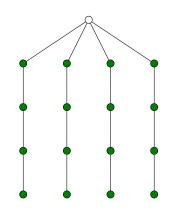
case 1: (1,4,2)

21 nodes

Randomly Selected Path 2

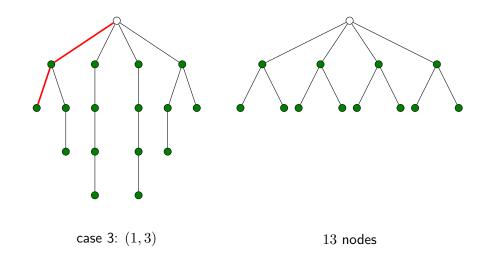


case 2: (2,4,1,3)



17 nodes

Randomly Selected Path 3



Estimation Result

Suppose sampling four times

- case 1: 1
- case 2: 1
- case 3: 2

Average number of nodes: $(21 \times 1 + 17 \times 1 + 13 \times 2)/4 = 16$

The real number of nodes: 17

 more samplings will make the estimation approaches the real number