

Lecture 5: Backtracking Algorithm

Analysis and Design of Computer Algorithms

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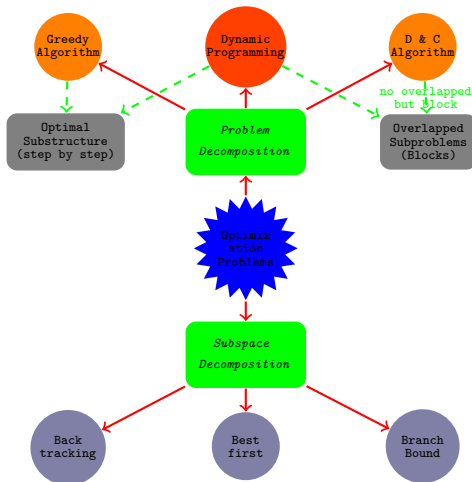
Website: <http://cs.tju.edu.cn/faculties/gongxj/course/algorithm>

Discussion : <http://groups.google.com/group/algorithm-practice-team>

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Backtracking Algorithm

Motivations



Solution space

- ▶ A solution to a specific problem can be represented by a n tuple $(x_1, x_2, \dots, x_n)$
 - ① x_i comes from a limited set S_i .
 - ② The dimensions of tuples for different solutions might vary depending on its definition, thus the representation can be classified:
 - ③ **Fixed Length Tuple (FLT)** representation: all tuples have same dimensions
 - ④ **Varied Length Tuple (VLT)** representation: tuples have different dimensions
- ▶ **Solution space**: A set consists of all enumerates in the form of solution representation.
 - ① For FLT: $X = \{(x_1, x_2, \dots, x_n) | x_i \in S_i\}$
 - ② For VLT: $X = \{X^1, X^2, \dots, X^k\}$ where $X^j = \{(x_{j1}, x_{j2}, \dots, x_{jk})\}$

Solution space, cont.

- ▶ Not all $x \in X$ is the real solution
 - ① **Feasible solution** : if x holds all constrain conditions
 $g_i(x) = \text{true}$ ($i \in [1..m]$), simply as $g(x)$
 - ② **Optimal solution** : if x maximize/minimize target functions
 $f_i(x)$ ($i \in [1..k]$), simply as $f(x)$
- ▶ Our goal is to search for the feasible/optimal solutions from solution space
 - ① Define solution representations
 - ② Formulate solution space
 - ③ Search solution space

Define solution representations

Eight queens puzzle

Eight queens puzzle

Using a regular chess board, the challenge is to place eight queens on the board such that no queen is attacking any of the others.

- ▶ The solution can be represented by a 8- tuple $(x_1, \dots, x_8)$ where x_i is the column number of i-th queen
- ▶ The size of solution space is 8^8
- ▶ Constrains: $x_i \neq x_j$ and $|x_i - x_j| \neq |j - i|$ for all i, j

	1	2	3	4	5	6	7	8
1				Q				
2						Q		
3								Q
4		Q						
5							Q	
6	Q							
7			Q					
8					Q			

Define solution representations

Subset sum problem

Subset sum problem

Finding what subset of a set of positive integers $S = \{w_1, w_2, \dots, w_n\}$ has a given sum M ,

For an example, $M=31$, $n=4$ and $W=(11, 13, 24, 7)$, then
 $11+13+7=31$ and $24+7=31$

► The solution can be represented by

FLT $\mapsto$	$(x_1, x_2, \dots, x_n)$ where $x_i = 1$ if it is chosen else 0	$(1,1,0,1)$ $(0,0,1,1)$	$O(2^n)$
VLT $\mapsto$	$(j_1, j_2, \dots, j_k)$ where j_i is the order number of i -th integer chosen in S and k is the total number chosen	$(1,2,4)$ $(3,4)$	$O(2^n)$

Formulate solution space

- ▶ **Problem state**: a point in solution subspace. We can check if it reaches the goal
- ▶ **Start state** / (root): a problem state at which we start to search for the goal
- ▶ **Solution state**: a path from current (visiting) state to the root
- ▶ **Answer/goal state**: a point in solution space that is the goal

Definition

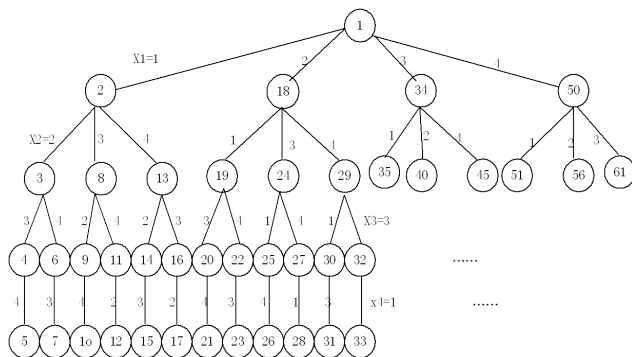
The **state space** of a problem is a 4-tuple (N, A, S, G) where:

- ▶ N is a set of problem states
- ▶ A is a set of arcs connecting the states
- ▶ S is a nonempty subset of N that contains start states
- ▶ G is a nonempty subset of N that contains the goal states.
- ▶ Our goal is to search solution states from S to G

State space tree

4-queens puzzle

State space tree is the representation of state space in the form of tree structures



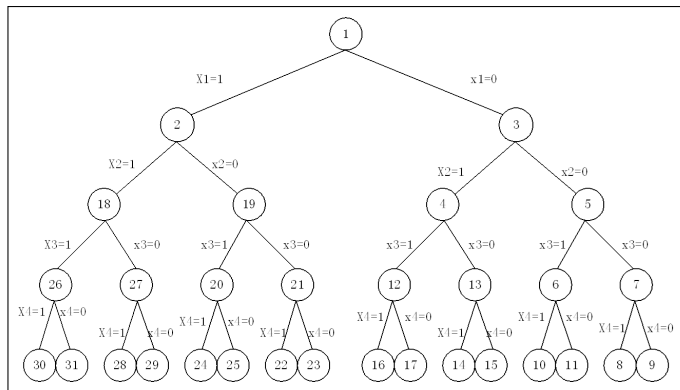
number inside a node is the order of depth first searching the tree
edge label is x_i and i is the depth of tree
This kind of tree is called Permutation Tree.

State space tree

Subset Sum

$M=31$, $n=4$, and $W=(11,13,24,7)$

For fixed length tuple representation

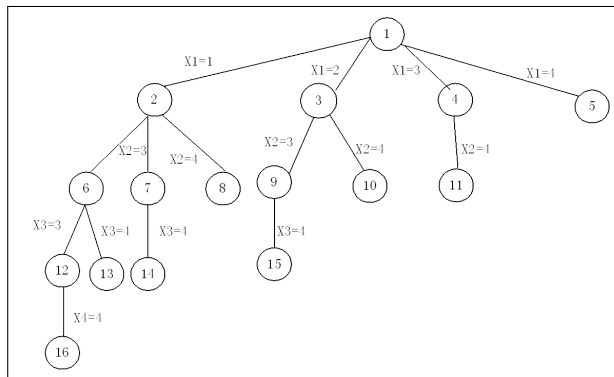


State space tree

Subset Sum

$M=31$, $n=4$, and $W=(11,13,24,7)$

For varied length tuple representation



Searching solution space

- ▶ Searching solutions equals to traverse the state space tree
- ▶ Node has three states during expending a tree
 - ① Live node: A node that has been reached, but not all of its children have been explored
 - ② Died node: A node where all of its children have been explored
 - ③ E-Node (expansion node) A live node in which its children are currently being explored.
- ▶ Search strategy
 - ① Backtrack: A variant version of depth first search with a bound function
 - ② Branch-bound(Best first search): an enhancement of backtracking

Bounding is a boolean function to kill a live node

Backtrack algorithm

Algorithm 1: Backtrack Algorithm

```

1 Function backtrack(int n)
2   k=1; while k > 0 do
3     forall the  $x[k] \in T(X(1), \dots, X(k-1))$  do
4       if not  $B(X(1), \dots, X(k))$  then
5         if  $(X(1), \dots, X(k))$  is an answer then
6            $\lfloor$  print  $(X(1), \dots, X(k))$  ;
7            $\lfloor$  k=k+1 /*loop next */
8         else
9            $\lfloor$  k =k-1 /*backtrack */

```

- ▶ $T(X(1), \dots, X(k-1))$ is a set containing all possible values $x(k)$, given $X(1), \dots, X(k-1)$
- ▶ $B(X(1), \dots, X(k))$ judge whether $X(k)$ satisfies constraints
- ▶ Solution is store in $X(1:n)$, once it is decided, output it

Bounding function for Subset sum problem

- ▶ Simple bounding: $B(X(1), \dots, X(k)) = \text{true}$ iff

$$\sum_{i=1}^k W(i)X(i) + \sum_{i=k+1}^n W(i) < M \quad (1)$$

- ▶ Tighter bounding: $B(X(1), \dots, X(k)) = \text{true}$ iff (1) and

$$\sum_{i=1}^k W(i)X(i) + W(k+1) > M \quad (2)$$

when sorting $W(i)$ by non-decreasing order

Subset sum algorithm with bounding

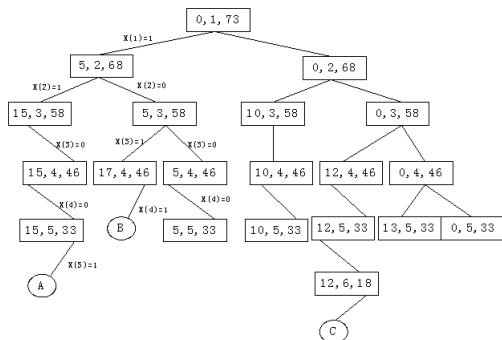
Algorithm 2: Subset sum problem: pseudo code

```
1 Let  $s = w(1)x(1) + \dots + w(k-1)x(k-1)$  ;  
2    $r = w(k) + \dots + w(n)$ , assumed  $s + r \geq M$  ;  
3 Expanding left child node ;  
4 if  $S + W(k) + W(k + 1) > M$  then  
5   | stop expanding ;  
6   |  $r = r - w(k)$  ;  
7   | Expanding right child node ;  
8 else  
9   |  $x(k) = 1$  ;  
10  |  $s = s + w(k)$  ;  
11  |  $r = r - w(k)$  ;  
12  | let  $(x(1), \dots, x(k))$  be E-Node ;  
13 Expanding right child node;  
14 if  $s + r < M$  or  $s + w(k + 1) > M$  then  
15   | stop expanding  
16 else  
17   |  $x(k) = 0$ 
```

State space tree with bounding

Subset sum problem

$M=30, n=6$ and $w=(5,10,12,13,15,18)$



Numbers in a rectangle node corresponds to s , k and r values respectively
 Circle nodes correspond to answer states
 There are only 23 nodes, but 63 nodes without bounding

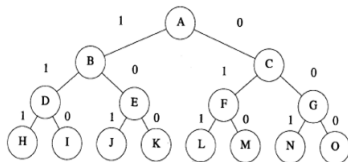
Problem statement

Container Loading Problem:CLP

- ▶ Given a ship has capacity c and n containers with weights $(w_1, \dots, w_n)$ are available for loading.
- ▶ Aiming at loading as many containers as is possible without sinking the ship.

Using fixed length tuple $x = (x_1, \dots, x_n)$ ($x_i = 1$, if container i is loaded) as solution space representation, an example of state space tree is shown below.

$n = 4, w = [8, 6, 2, 3], c = 12$



Bound function

- Suppose node $i-1$ is the E-node, let

$$cw = \sum_{j=1}^{i-1} x_j w_j$$

total weight of containers
loaded already

$$r = \sum_{j=i}^n w_j$$

total weight of unloaded
containers

$$bestw =$$

the optimal total weight
up to now

- 1 $\text{Bound1}(x_1, \dots, x_i) = \text{true}$ if $cw + w_i > c$: Kill node i
- 2 $\text{Bound2}(x_1, \dots, x_i) = \text{true}$ $cw + r \leq bestw$: stop expanding node i .
- 3 Above two bounding functions can be used at same time

Backtrack algorithm for CLP

```

1  template<class T>
2  T MaxLoading(T w[], T c, int n, int
      bestx[])
3  {// Return best loading and its
      value.
4      Loading<T> X;
5      // initialize X
6      X.x = new int [n+1];
7      X.w = w;
8      X.c = c;
9      X.n = n;
10     X.bestx = bestx;
11     X.bestw = 0;
12     X.cw = 0;
13     // initial r is sum of all
       weights
14     X.r = 0;
15     for (int i = 1; i <= n; i++)
16         X.r += w[i];
17     X.maxLoading(1);
18     delete [] X.x;
19     return X.bestw;
20 }

```

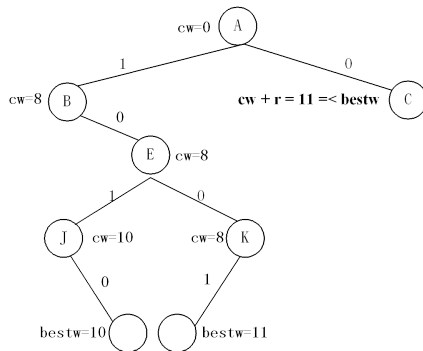
```

1  template<class T>
2  void Loading<T>::maxLoading(int i)
3  {// Search from level i node.
4      if (i > n) {// at a leaf
5          for (int j = 1; j <= n; j++)
6              bestx[j] = x[j];
7          bestw = cw; return;}
8      // check subtrees
9      r -= w[i];
10     if (cw + w[i] <= c) {// try x[i]
11         = 1
12         x[i] = 1;
13         cw += w[i];
14         maxLoading(i+1);
15         cw -= w[i];}
16     if (cw + r > bestw) {// try x[i]
17         = 0
18         x[i] = 0;
19         maxLoading(i+1);}
20     r += w[i];
21 }

```

State space tree with bounding

$$n = 4, w = [8, 6, 2, 3], c = 12$$



Bound function for 0/1 Knapsack

- ▶ Suppose that items are sorted in non-decreasing order of p/w and node $k-1$ is the E-node, let

$$cp = \sum_{j=1}^{k-1} x_j p_j \quad \text{profit of current packing}$$

$$rp = \sum_{j=k}^n p_j \quad \text{total profit of remain items}$$

$$bestp = \quad \text{max profit so far}$$

- ▶ $\text{Bound}(x) = \text{true}$ if $cp + rp \leq bestp$

Backtrack algorithm for Knapsack

```

1  template<class Tw, class Tp>
2  void Knap<Tw, Tp>::Knapsack(int i)
3  {// Search from level i node.
4      if (i > n) {// at a leaf
5          bestp = cp;
6          return;}
7      // check subtrees
8      if (cw + w[i] <= c) {// try x[i]
9          = 1
10         cw += w[i];
11         cp += p[i];
12         Knapsack(i+1);
13         cw -= w[i];
14         cp -= p[i];}
15     if (Bound(i+1) > bestp) // try x[i]
16         = 0
17         Knapsack(i+1);
18 }

```

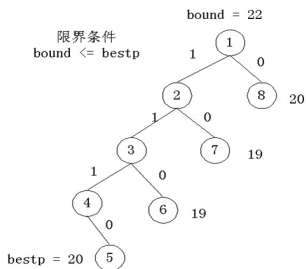
```

1  template<class Tw, class Tp>
2  Tp Knap<Tw, Tp>::Bound(int i)
3  {// Return upper bound on value of
4      // best leaf in subtree.
5      Tw cleft = c - cw; // remaining
6          capacity
7      Tp b = cp; // profit
8          bound
9      // fill remaining capacity
10     // in order of profit density
11     while (i <= n && w[i] <= cleft) {
12         cleft -= w[i];
13         b += p[i];
14         i++;
15     }
16     // take fraction of next object
17     if (i <= n) b += p[i]/w[i] *
18         cleft;
19     return b;
20 }

```

State space tree with bounding

$$n = 4, c = 7, p = [9, 10, 7, 4], w = [3, 5, 2, 1]$$



Max Clique: problem statements

A subgraph $G' = \langle V', E' \rangle$ is a **complete subgraph** of an undirected graph $G = \langle V, E \rangle$, if and only if $V' \subset V$ and for $\forall u \in V', \forall v \in V', (u, v) \in E' \subset E$.

A **clique** is a complete subgraph of G if no larger inclusion of other complete subgraphs.

A **max clique** is a clique of the largest possible size in a given graph.

A **independent vertex set** is a subgraph of G with empty edges if no larger inclusion of other independent vertex sets.

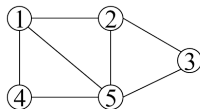
A **max independent vertex set** is a independent vertex set of the largest possible size in a given graph.

An example

$\{1,2\}$ is a complete subgraph, but not a clique

$\{1,2,5\}$ $\{1,4,5\}$ $\{2,5,5\}$ are max cliques

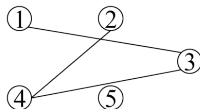
$\{2,4\}$ is a max independent vertex set



$\{1,2\}$ is a empty subgraph, but not a independent vertex set

$\{2,3\}$ $\{1,2,5\}$ are independent vertex sets

$\{1,2,5\}$ is also a max independent vertex sets



Max clique: bounding

- ▶ Our goal is to find the max cliques of given graph G
- ▶ Using fixed length tuple $X=x[1..n]$ ($x[i]=1$ if vertex i is included) to represent solution space
- ▶ Its state space tree is a subset tree
- ▶ Bounding
 - ① $B(x)=\text{true}$ if the vertexes from root to i can not form a complete subgraph
 - ② $B(x)=\text{true}$ if the number of vertexes from root to i plus remained vertexes is no larger than bestn

Backtrack algorithm for maxclique

```

1  int AdjacencyGraph::MaxClique(
2      int v[])
3  {// Return size of largest
4      clique.
5      // Return clique vertices in v
6      [1:n].
7      // initialize for maxClique
8      x = new int [n+1];
9      cn = 0;
10     bestn = 0;
11     bestx = v;
12
13     // find max clique
14     maxClique(1);
15
16     delete [] x;
17     return bestn;
18 }

```

```

1  void AdjacencyGraph::maxClique(int i)
2  {// Backtracking code to compute largest cliq
3      if (i > n) {// at leaf
4          // found a larger clique, update
5          for (int j = 1; j <= n; j++)
6              bestx[j] = x[j];
7          bestn = cn;
8          return;}
9      // see if vertex i connected to others
10     // in current clique
11     int OK = 1;
12     for (int j = 1; j < i; j++)
13         if (x[j] && a[i][j] == NoEdge) {
14             // i not connected to j
15             OK = 0;
16             break;}
17
18     if (OK) {// try x[i] = 1
19         x[i] = 1; // add i to clique
20         cn++;
21         maxClique(i+1);
22         x[i] = 0;
23         cn--;}
24
25     if (cn + n - i > bestn) {// try x[i] = 0
26         x[i] = 0;
27         maxClique(i+1);}
28 }

```

TSP

- ▶ Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city **exactly once** and returns to the **origin city**?
- ▶ It can be modeled as an **undirected weighted graph**, such that cities are the graph's vertexes, paths are the graph's edges, and a path's distance is the edge's length.
- ▶ It is a **minimization problem** starting and finishing at a specified vertex after having visited each other vertex exactly once.

Bounding

- ▶ Define $x = x[1..n]$ (x_i is the order number of i -th vertex in the route) as the solution representation
- ▶ Its state space tree is a permutation tree
- ▶ Bounding
 - ① $B(i) = \text{true}$ if no edge connection between $x[i]$ and $x[i-1]$
 - ② $B(i) = \text{true}$ if route distance from root to $x[i]$ is larger than bestc (bestc is the shortest route distance so far)

Backtrack algorithm for maxclique

```

1  void AdjacencyWDigraph<T>::tSP(int i)
2  {// Backtracking code for traveling salesperson.
3      if (i == n) {// at parent of a leaf
4          // complete tour by adding last two edges
5          if (a[x[n-1]][x[n]] != NoEdge &&
6              a[x[n]][1] != NoEdge &&
7              (cc + a[x[n-1]][x[n]] + a[x[n]][1] <
                bestc ||
8              bestc == NoEdge)) {// better tour found
9              for (int j = 1; j <= n; j++)
10                  bestx[j] = x[j];
11              bestc = cc + a[x[n-1]][x[n]] + a[x[n]
                ][1];}
12      }
13      else {// try out subtrees
14          for (int j = i; j <= n; j++)
15              // is move to subtree labeled x[j]
                possible?
16              if (a[x[i-1]][x[j]] != NoEdge &&
17                  (cc + a[x[i-1]][x[i]] < bestc ||
18                  bestc == NoEdge)) {// yes
19                  // search this subtree
20                  Swap(x[i], x[j]);
21                  cc += a[x[i-1]][x[i]];
22                  tSP(i+1);
23                  cc -= a[x[i-1]][x[i]];
24                  Swap(x[i], x[j]);}
25      }
26  }
```

```

1  template<class T>
2  T AdjacencyWDigraph<T>::TSP(int i,
3      v[])
4  {// Traveling salesperson by
        backtracking.
5      // Return cost of best tour,
        return tour in v[1:n].
6      // initialize for tSP
7      x = new int [n+1];
8      // x is identity permutati
9      for (int i = 1; i <= n; i++)
10          x[i] = i;
11      bestc = NoEdge;
12      bestx = v; // use array v
        store best tour
13      cc = 0;
14
15      // search permutations of
        [2:n]
16      tSP(2);
17
18      delete [] x;
19      return bestc;
20  }
```