#### **CS-258**

# Solving Hard Problems in Combinatorial Optimization

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## **Course Overview**

## Study of combinatorial optimization from a practitioner standpoint

- the main techniques and tools
- several representative problems

## What is expected from you:

- 7 team assignments (2 persons team)
- Short write-up on the assignments
- Final presentation of the results

## What you should not do in this class?

- read papers/books/web/... on combinatorial optimization
- Read the next slide during class

## **Course Overview**

```
January 31: Knapsack
     * Assignment: Coloring
February 7: LS + CP + LP + IP
February 14: Linear Programming
     * Assignment: Supply Chain
February 28: Integer Programming
     * Assignment: Green Zone
March 7: Travelling Salesman Problem
     * Assignment: TSP
March 14: Advanced Mathematical Programming
March 21: Constraint Programming
     * Assignment: Nuclear Reaction
April 4: Constraint-Based Scheduling
April 11: Approximation Algorithms (Claire)
     * Assignment: Operation Hurricane
April 19: Vehicle Routing
April 25: Advanced Constraint Programming
     * Assignment: Special Ops.
```

## Knapsack Problem

$$\max \sum_{i=1}^{n} c_i x_i$$

## subject to:

$$\sum_{j=1}^{n} W_{j} X_{j} \leq I$$

$$x_i \in [0, 1]$$

## NP-Complete Problems

#### Decision problems

#### Polynomial-time verifiability

• We can verify if a guess is a solution in polynomial time

## Hardest problem in NP (NP-complete)

• If I solve one NP-complete problem, I solve them all

## **Approaches**

#### Exponential Algorithms (sacrificing time)

- dynamic programming
- branch and bound
- branch and cut
- constraint programming

## Approximation Algorithms (sacrificing quality)

- polynomial time
- guarantee of certain performance

## Local Search Algorithms (sacrificing quality)

- iterative improvement of a solution
- not always a performance guarantee

$$\max \sum_{i=1}^{n} c_i x_i$$

## subject to:

$$\sum_{j=1}^{n} W_j X_j \le I \qquad x_j \in [0, 1]$$

#### Divide and Conquer:

P(k, d) is defined as:  $\max_{i=1}^{\infty} \sum_{j=1}^{\infty} c_j x_j$ 

subject to 
$$\sum_{j=1}^{k} W_j X_j \le d$$

variable 1 ... k capacity is d

Solve: P(k, d)

• take variable k  $(x_k = 1)$ 

$$\max \sum_{i=1}^{k-1} c_i x_i + c_k$$

subject to 
$$\sum_{i=1}^{k-1} w_i x_i \le d - w_k$$

• drop variable  $k (x_k = 0)$ 

max 
$$\sum_{i=1}^{k-1} c_i x_i$$
subject to 
$$\sum_{i=1}^{k-1} W_i x_i \le d$$

#### Original Problem:

Basic Divide and Conquer step for p(k, d)

• take variable k if  $W_k \le d$ 

$$c_k + p((k-1),(d-w_k))$$

• drop variable k

$$p(k-1,d)$$

$$p(k, d) = max(c_k + p(k-1, d-w_k), p(k-1, d))$$

$$p(0, d) = 0$$

#### Recursive formulation

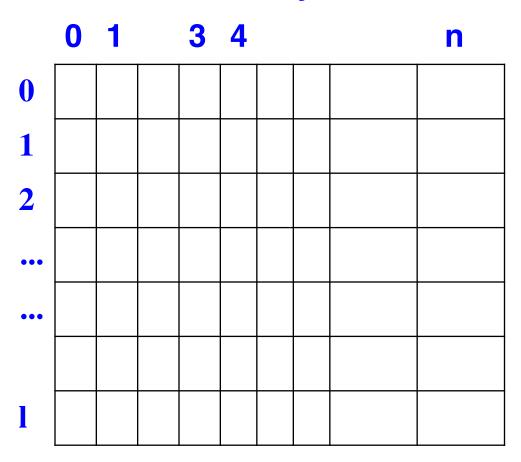
```
long p(k, d) {
   if k = 0 then
      return 0;
   else if w<sub>k</sub> > d then
      return p(k-1, d);
   else
      return max(p(k-1, d), ck + p(k-1, d-w<sub>k</sub>))
   }
```

Is this algorithm any good?

```
long fib(n) {
  if n = 0 then return 1;
  else if n = 1 then return 1;
  else return fib(n-1) + fib(n-2);
}
```

Is this any good?

# of objects



```
for i := 1 to n do
  for d := 0 to l do
    if w<sub>i</sub> > d then
       p[d, i] = p[d, i-1]
    else
       p[d, i] = max(c<sub>i</sub> + p[d-w<sub>i</sub>, i-1],p[d, i-1])
```

Complexity ??

Complexity: O(ln)

Is this polynomial contradicting  $P \neq NP$ ?

How do you represent 1 on a computer?

Pseudo Polynomial Algorithm

 Polynomial algorithm if the numbers are small

## Knapsack: Example

max 
$$16x_1 + 19x_2 + 23x_3 + 28x_4$$
  
subject to  $2x_1 + 3x_2 + 4x_3 + 5x_4 \le 7$   
 $x_i \in [0, 1]$ 

	0	1	2	3	4
0	0 🖈	0	0	0	0
1	0	0	0	0	0
2	0	<sup>1</sup> 16 ←	<u>     16 </u> <b>←</b>	<u> </u>	16
3	0	16	19	19 \	19
4	0	16	19	23	23
5	0	16	35	35	35
6	0	16	35	39	39
7	0	16	35	42	<b>\</b> 44

How do we get the optimal solution?

$$X_n^* = 0$$
 if  $p(n, l) = p(n-1, l)$   
 $X_n^* = 1$  otherwise

Denote **d(k)** = 1 - 
$$\sum_{i=k+1}^{n} w_i x_i^*$$

We have

$$X^*k = 0$$
 if  $p(n, d(k)) = p(n-1, d(k))$ 
 $X^*k = 1$  otherwise.

 $d := 1$ 

for  $k := n$  down to 1 do

if  $p[d, k] = p[d, k-1]$  then

 $X^*k = 0$ ;

else

 $X^*k = 1$ ;
 $d := d - w_k$ ;

## Generalizations

**Q:** Can you generalize it if the variables are not bounded?

**Q:** Can we solve the problem in linear space in l?

## Knapsack Generalization

$$\max \sum_{i=1}^{n} c_i x_i \quad \text{subject to} \quad \sum_{i=1}^{n} W_i x_i \le L$$

Define p(d) as

$$\max \sum_{i=1}^{n} c_i x_i \quad \text{subject to} \quad \sum_{i=1}^{n} w_i x_i \le d$$

Now

$$p(d) = \max\{c_j + p(d-w_j) \mid w_j \le d \& 1 \le j \le n\}$$

Complexity

O(nL)

One of the most successful techniques for solving optimization problems optimally

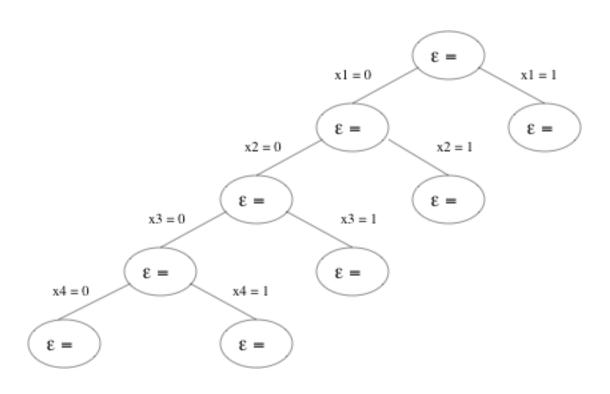
Divide and Conquer

#### Two Steps

- **Bounding:** getting an optimistic estimation of the optimum for a subproblem (key step)
- **Branching:** split the problem into subproblems (key step)

Why is bounding so important?

max 
$$16x_1 + 19x_2 + 23x_3 + 28x_4$$
  
subject to  $2x_1 + 3x_2 + 4x_3 + 5x_4 \le 7$   
 $x_i \in [0, 1]$ 



#### Best-first:

- explore the node with the best evaluation function
- when does pruning occur?
- what is the main limitation?

## Depth first:

when does pruning occur?

## There are other strategies as well

• Limited Discrepancy Search

#### How to bound?

Relaxation

#### How to relax?

Remove some constraints

## Knapsack Problem

$$\max \sum_{i=1}^{n} c_i x_i$$

## subject to:

$$\sum_{j=1}^{n} W_{j} X_{j} \leq I$$

$$x_i \in [0, 1]$$

#### Naïve evaluation

• remove the capacity constraint

$$\max c_1 v_1 + ... + c_m v_m + \sum_{i=m+1}^{n} c_i x_i$$

subject to 
$$x_i \in [0, 1]$$

#### Linear Programming Relaxation

• remove the integrality requirement

$$\max \sum_{i=1}^{n} c_i x_i$$

subject to 
$$\sum_{i=1}^{n} W_i X_i \le I \qquad 0 \le X \le 1$$

What is the optimal solution assuming that

$$\frac{c_1}{W_1} \ge \frac{c_2}{W_2} \ge \dots \ge \frac{c_n}{W_n}$$

$$\max \sum_{i=1}^{n} c_i x_i$$

subject to 
$$\sum_{j=1}^{n} w_j x_j \le l \qquad 0 \le x \le 1$$

Rewrite 
$$x_i = \frac{y_i}{c_i}$$
 to obtain

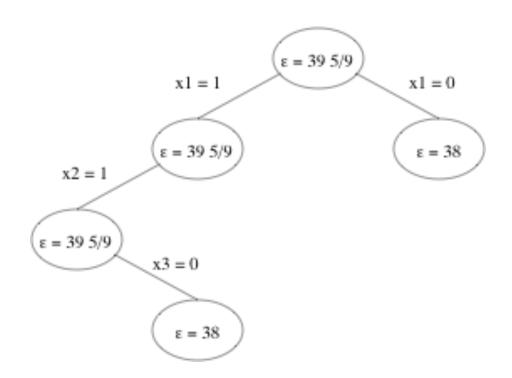
$$\max \sum_{i=1}^{n} y_i$$

subject to 
$$\sum_{j=1}^{n} \frac{w_j}{c_j} y_j \le I \qquad y_i \in [0...1]$$

We have that 
$$\frac{w_1}{c_1} \le \frac{w_2}{c_2} \le \dots \le \frac{w_n}{c_n}$$

What is the optimal solution?

$$3x_1 + 7x_2 + 9x_3 + 6x_4 + 3x_5 + 5x_6 + 2x_7 \le 17$$



## Time to Pack

**Q:** Can you think of a way to combine both algorithms?