

DM811
Heuristics for Combinatorial Optimization

Lecture 16
Examples

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Outline

1. Recap.
2. Other Combinatorial Optimization Problems
 - Quadratic Assignment Problem
 - School Scheduling
 - Linear Ordering
 - Bin Packing
3. Appendix
 - Random Numbers

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Summary: Local Search Algorithms

(as in [Hoos, Stützle, 2005])

For given problem instance π :

1. search space S_π
2. neighborhood relation $\mathcal{N}_\pi \subseteq S_\pi \times S_\pi$
3. evaluation function $f_\pi : S \rightarrow \mathbf{R}$
4. set of memory states M_π
5. initialization function $\text{init} : \emptyset \rightarrow S_\pi \times M_\pi$
6. step function $\text{step} : S_\pi \times M_\pi \rightarrow S_\pi \times M_\pi$
7. termination predicate $\text{terminate} : S_\pi \times M_\pi \rightarrow \{\top, \perp\}$

After implementation and test of the above components, improvements in efficiency (ie, computation time) can be achieved by:

- A. fast delta evaluation
- B. neighborhood pruning
- C. clever use of data structures

Improvements in quality can be achieved by:

- D. application of a metaheuristic
- E. definition of a larger neighborhood

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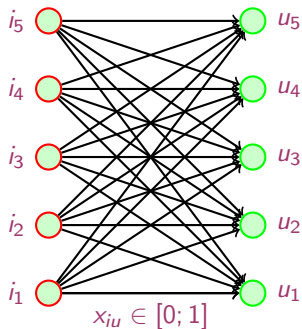
Quadratic Assignment Problem

- **Given:**
 n units with a matrix $F = [f_{ij}] \in \mathbf{R}^{n \times n}$ of flows between them and
 n locations with a matrix $D = [d_{uv}] \in \mathbf{R}^{n \times n}$ of distances
- **Task:** Find the assignment σ of units to locations that minimizes the sum of product between flows and distances, ie,

$$\min_{\sigma \in \Sigma} \sum_{i,j} f_{ij} d_{\sigma(i)\sigma(j)}$$

Applications: hospital layout; keyboard layout

Quadratic Programming Formulation



indices i, j for units and u, v for locations:

$$\begin{aligned} \min \quad & \sum_i \sum_u \sum_j \sum_v f_{ij} d_{uv} x_{iu} x_{jv} + \left(\sum_i \sum_u c_{iu} x_{iu} \right) \\ \text{s.t.} \quad & \sum_i x_{iu} = 1 \quad \forall u \\ & \sum_u x_{iu} = 1 \quad \forall i \\ & x \geq 0 \text{ and integer } \forall i, u \end{aligned}$$

Largest instances solvable exactly $n = 30$

Example: QAP

$$D = \begin{pmatrix} 0 & 4 & 3 & 2 & 1 \\ 4 & 0 & 3 & 2 & 1 \\ 3 & 3 & 0 & 2 & 1 \\ 2 & 2 & 2 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \end{pmatrix} \quad F = \begin{pmatrix} 0 & 1 & 2 & 3 & 4 \\ 1 & 0 & 2 & 3 & 4 \\ 2 & 2 & 0 & 3 & 4 \\ 3 & 3 & 3 & 0 & 4 \\ 4 & 4 & 4 & 4 & 0 \end{pmatrix}$$

The optimal solution is $\sigma = (1, 2, 3, 4, 5)$, that is,
facility 1 is assigned to location 1,
facility 2 is assigned to location 2, etc.

The value of $f(\sigma)$ is 100.

Delta evaluation

Evaluation of 2-exchange $\{r, s\}$ can be done in $O(n)$

$$\begin{aligned} \Delta(\psi, r, s) = & b_{rr} \cdot (a_{\psi_s \psi_s} - a_{\psi_r \psi_r}) + b_{rs} \cdot (a_{\psi_s \psi_r} - a_{\psi_r \psi_s}) + \\ & b_{sr} \cdot (a_{\psi_r \psi_s} - a_{\psi_s \psi_r}) + b_{ss} \cdot (a_{\psi_r \psi_r} - a_{\psi_s \psi_s}) + \\ & \sum_{k=1, k \neq r, s}^n (b_{kr} \cdot (a_{\psi_k \psi_s} - a_{\psi_k \psi_r}) + b_{ks} \cdot (a_{\psi_k \psi_r} - a_{\psi_k \psi_s}) + \\ & b_{rk} \cdot (a_{\psi_s \psi_k} - a_{\psi_r \psi_k}) + b_{sk} \cdot (a_{\psi_r \psi_k} - a_{\psi_s \psi_k})) \end{aligned}$$

Example: Tabu Search for QAP

- **Solution representation:** permutation π
- **Initial Solution:** randomly generated
- **Neighborhood:** interchange
 $\Delta_I : \delta(\pi) = \{\pi' | \pi'_k = \pi_k \text{ for all } k \neq \{i, j\} \text{ and } \pi'_i = \pi_j, \pi'_j = \pi_i\}$
- **Tabu status:** forbid δ that place back the items in the positions they have already occupied in the last tt iterations (short term memory)
- Implementation details:
 - compute $g(\pi') - f(\pi)$ in $O(n)$ or $O(1)$ by storing the values all possible previous moves.
 - maintain a matrix $[T_{ij}]$ of size $n \times n$ and write the last time item i was moved in location k plus tt
 - δ is tabu if it satisfies both:
 - $T_{i, \pi(j)} \geq$ current iteration
 - $T_{j, \pi(i)} \geq$ current iteration

Example: Robust Tabu Search for QAP

- **Aspiration criteria:**
 - allow forbidden δ if it improves the last π^*
 - select δ if never chosen in the last A iterations (long term memory)
- Parameters: $tt \in [[0.9n], [1.1n + 4]]$ and $A = 5n^2$

Example: Reactive Tabu Search for QAP

- **Aspiration criteria:**

- allow forbidden δ if it improves the last π^*

- **Tabu Tenure**

- maintain a hash table (or function) to record previously visited solutions
- increase tt by a factor $\alpha_{inc}(= 1.1)$ if the current solution was previously visited
- decrease tt by a factor $\alpha_{dec}(= 0.9)$ if tt not changed in the last $sttc$ iterations or all moves are tabu
- Trigger escape mechanism if a solution is visited more than $nr(= 3)$ times
- Escape mechanism = $1 + (1 + r) \cdot ma/2$ random moves

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School scheduling

Input: a finite set of **time periods** and **courses** with assigned: a teacher, a set of attending students and a suitable room.

Task: Produce weekly **timetable** of courses, that is: assign a time period of the week (typically one hour) to every course such that courses are assigned to different time periods if:

- they are taught by the same teacher
- they can be held only in the same room
- they share students.

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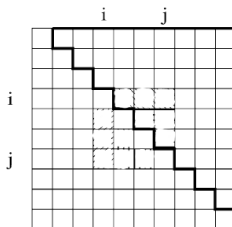
Linear Ordering Problem

Input: an $n \times n$ matrix C

Task: Find a permutation π of the column and row indices $\{1, \dots, n\}$ such that the value

$$f(\pi) = \sum_{i=1}^n \sum_{j=i+1}^n c_{\pi_i \pi_j}$$

is maximized. In other terms, find a permutation of the columns and rows of C such that the sum of the elements in the upper triangle is maximized.



Consider as an example the (5,5)-matrix:

$$H = \begin{bmatrix} 0 & 16 & 11 & 15 & 7 \\ 21 & 0 & 14 & 15 & 9 \\ 26 & 23 & 0 & 26 & 12 \\ 22 & 22 & 11 & 0 & 13 \\ 30 & 28 & 25 & 24 & 0 \end{bmatrix}$$

$\pi = (1, 2, 3, 4, 5)$. The sum of its superdiagonal elements is 138.

$\pi = (5, 3, 4, 2, 1)$ i.e., H_{12} becomes $H_{\pi(1)\pi(2)} = H_{54}$ in the permuted matrix.

Thus the optimal triangulation of H is

$$H^* = \begin{bmatrix} 0 & 25 & 24 & 28 & 30 \\ 12 & 0 & 26 & 23 & 26 \\ 13 & 11 & 0 & 22 & 22 \\ 9 & 14 & 15 & 0 & 21 \\ 7 & 11 & 15 & 16 & 0 \end{bmatrix}$$

Now the sum of superdiagonal elements is 247.

LOP Applications: Graph Theory

Definition: A **directed graph** (or **digraph**) D consists of a non-empty finite set $V(D)$ of distinct vertices and a finite set A of ordered pairs of distinct vertices called arcs.

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Feedback arc set problem (FASP)

Input: A directed graph $D = (V, A)$, where $V = \{1, 2, \dots, n\}$, and arc weights c_{ij} for all $[ij] \in A$

Task: Find a permutation $\pi_1, \pi_2, \dots, \pi_n$ of V (that is, a linear ordering of V) such that the total costs of those arcs $[\pi_j \pi_i]$ where $j > i$ (that is, the arcs that point backwards in the ordering)

$$f(\pi) = \sum_{i=1}^n \sum_{j=i+1}^n c_{\pi_j \pi_i}$$

is minimized.

LOP Applications: Graph Theory (2)

Definition: A **linear ordering** of a finite set of vertices $V = \{1, 2, \dots, n\}$ is a bijective mapping (permutation) $\pi : \{1, 2, \dots, n\} \rightarrow V$. For $u, v \in V$, we say that u is “before” v if $\pi^{-1}(u) < \pi^{-1}(v)$ ($\pi^{-1}(i) = \text{pos}_\pi(i)$).

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Definition: An **oriented** graph is a digraph with no cycle of length two. A **tournament** is an oriented graph where every pair of distinct vertices are adjacent.

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Remark: Given a digraph $D = (V, A)$ and a linear ordering of the vertices V , the arc set $E = \{[uv] \mid \pi^{-1}(u) < \pi^{-1}(v)\}$ forms an acyclic tournament on V . Similarly, an acyclic tournament $T = (V, E)$ induces a linear ordering of V .

Definition: The cost of a linear ordering is expressed by

$$\sum_{\pi^{-1}(u) < \pi^{-1}(v)} c_{uv}$$

where the costs c_{uv} are the costs associated to the arcs.

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Linear Ordering Problem

Input: Given a complete digraph $D = (V, A)$ with arc weights c_{ij} for all $ij \in A$

Task: Find an acyclic tournament $T = (V, T)$ in D such that

$$f(T) = \sum_{ij \in T} c_{ij}$$

is maximized.

Aggregation of Individual Preferences

Kemeny's problem. Suppose that there are m persons and each person i , $i = 1, \dots, m$, has ranked n objects by giving a linear ordering T_i of the objects. Which common linear ordering aggregates the individual orderings in the best possible way?

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\rightsquigarrow linear ordering problem by setting c_{ij} = number of persons preferring object O_i to object O_j

LOP Applications: Economics

Input-output analysis (Leontief, Nobel prize)

The economy of a state is divided into n sectors, and an $n \times n$ input-output matrix C is constructed where the entry c_{ij} denotes the transactions from sector i to sector j in that year.

Triangulation (ie, solving associated LOP) allows identification of important inter-industry relations in an economy (from primary stage sectors via the manufacturing sectors to the sectors of final demand) and consequent comparisons between different countries.

Depicts dependencies between the different branches of an economy

Ranking in Sports Tournaments

H_{ij} = number of goals which were scored by team i against team j .

Table 1.1 Premier League 2006/2007 (left: official, right: triangulated)

1 Manchester United	1 Chelsea
2 Chelsea	2 Arsenal
3 Liverpool	3 Manchester United
4 Arsenal	4 Everton
5 Tottenham Hotspur	5 Portsmouth
6 Everton	6 Liverpool
7 Bolton Wanderers	7 Reading
8 Reading	8 Tottenham Hotspur
9 Portsmouth	9 Aston Villa
10 Blackburn Rovers	10 Blackburn Rovers
11 Aston Villa	11 Middlesbrough
12 Middlesbrough	12 Charlton Athletic
13 Newcastle United	13 Bolton Wanderers
14 Manchester City	14 Wigan Athletic
15 West Ham United	15 Manchester City
16 Fulham	16 Sheffield United
17 Wigan Athletic	17 Fulham
18 Sheffield United	18 Newcastle United
19 Charlton Athletic	19 Watford
20 Watford	20 West Ham United

R. Martí, G. Reinelt, R. Martí and G. Reinelt. *The Linear Ordering Problem, Introduction*. Springer Berlin Heidelberg, 2011, 1-15

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Knapsack, Bin Packing, Cutting Stock

Knapsack

Given: a knapsack with maximum weight W and a set of n items $\{1, 2, \dots, n\}$, with each item j associated to a profit p_j and to a weight w_j .

Task: Find the subset of items of maximal total profit and whose total weight is not greater than W .

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One dimensional Bin Packing

Given: A set $L = (a_1, a_2, \dots, a_n)$ of items, each with a size $s(a_i) \in (0, 1]$ and an unlimited number of unit-capacity bins B_1, B_2, \dots, B_m .

Task: Pack all the items into a minimum number of unit-capacity bins B_1, B_2, \dots, B_m .

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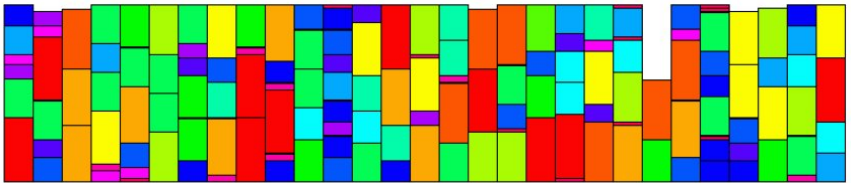
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Cutting stock

Each item has a profit $p_j > 0$ and a number of times it must appear a_j . The task is to select a subset of items to be packed in a single finite bin that maximizes the total selected profit.

Bin Packing



Cutting Stock

	0	1120	2240	3360	4480	5600 mm
2x ▶		1820	1820	1820		
3x ▶		1380	2150	1930		
12x ▶		1380	2150	2050		
7x ▶		1380	2100	2100		
12x ▶		2200	1820	1560		
8x ▶		2200	1520	1880		
1x ▶		1520	1930	2150		
16x ▶		1520	1930	2140		
10x ▶		1710	2000	1880		
2x ▶		1710	1710	2150		

Heuristics for Bin Packing

- Construction Heuristics
 - Best Fit Decreasing (BFD)
 - First Fit Decreasing (FFD)

$$C_{max}(FFD) \leq \frac{11}{9} C_{max}(OPT) + \frac{6}{9}$$

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- Local Search: [Alvim and Aloise and Glover and Ribeiro, 1999]
 - Step 1: remove one bin and redistribute items by BFD
 - Step 2: if infeasible, re-make feasible by redistributing items for pairs of bins, such that their total weights becomes equal (number partitioning problem)

[Levine and Ducatelle, 2004]

The solution before local search (the bin capacity is 10):

The bins: | 3 3 3 | 6 2 1 | 5 2 | 4 3 | 7 2 | 5 4 |

Open the two smallest bins:

Remaining: | 3 3 3 | 6 2 1 | 7 2 | 5 4 |

Free items: 5, 4, 3, 2

Try to replace 2 current items by 2 free items, 2 current by 1 free or 1 current by 1 free:

First bin: 3 3 3 → 3 5 2 new free: 4, 3, 3, 3

Second bin: 6 2 1 → 6 4 new free: 3, 3, 3, 2, 1

Third bin: 7 2 → 7 3 new free: 3, 3, 2, 2, 1

Fourth bin: 5 4 stays the same

Reinsert the free items using FFD:

Fourth bin: 5 4 → 5 4 1

Make new bin: 3 3 2 2

Final solution: | 3 5 2 | 6 4 | 7 3 | 5 4 1 | 3 3 2 2 |

Repeat the procedure: no further improvement possible

Two-Dimensional Packing Problems

Two dimensional bin packing

Given: A set $L = (a_1, a_2, \dots, a_n)$ of n rectangular *items*, each with a width w_j and a height h_j and an unlimited number of identical rectangular bins of width W and height H .

Task: Allocate all the items into a minimum number of bins, such that the original orientation is respected (no rotation of the items is allowed).

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Two dimensional strip packing

Given: A set $L = (a_1, a_2, \dots, a_n)$ of n rectangular *items*, each with a width w_j and a height h_j and a bin of width W and infinite height (*a strip*).

Task: Allocate all the items into the strip by minimizing the used height and such that the original orientation is respected (no rotation of the items is allowed).

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Two dimensional cutting stock

Each item has a profit $p_j > 0$ and the task is to select a subset of items to be packed in a single finite bin that maximizes the total selected profit.

Three dimensional

Given: A set $L = (a_1, a_2, \dots, a_n)$ of rectangular boxes, each with a width w_j , height h_j and depth d_j and an unlimited number of three-dimensional bins B_1, B_2, \dots, B_m of width W , height H , and depth D .

Task: Pack all the boxes into a minimum number of bins, such that the original orientation is respected (no rotation of the boxes is allowed)

List of Problems

See <http://www.nada.kth.se/~viggo/problemelist/>

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- long period
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- uncorrelated (time series analysis)
- efficient

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Suggested: MRG32k3a by L'Ecuyer

<http://www.iro.umontreal.ca/~lecuyer/>

```
java.lang.Object
```

```
  extended by umontreal.iro.lecuyer.rng.RandomStreamBase
```

```
    extended by umontreal.iro.lecuyer.rng.MRG32k3a
```


Ideal Random Shuffle

Let's consider a sequence of n elements: $\{e_1, e_2, \dots, e_n\}$.

The **ideal random shuffle** is a permutation chosen uniformly at random from the set of all possible $n!$ permutations.

- π_1 is uniformly randomly chosen among $\{e_1, e_2, \dots, e_n\}$.
- π_2 is uniformly randomly chosen among $\{e_1, e_2, \dots, e_n\} - \{\pi_1\}$.
- π_3 is uniformly randomly chosen among $\{e_1, e_2, \dots, e_n\} - \{\pi_1, \pi_2\}$
- ...

Joint probability of $(\pi_1, \pi_2 \dots \pi_n)$ is $\frac{1}{n} \cdot \frac{1}{n-1} \cdot \dots \cdot 1 = \frac{1}{n!}$

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- ...

Joint probability of $(\pi_1, \pi_2 \dots \pi_n)$ is $\frac{1}{n} \cdot \frac{1}{n-1} \cdot \dots \cdot 1 = \frac{1}{n!}$

```
long int* Random::generate_random_array(const int& size) {
    long int i, j, help;
    long int *v = new long int[size];
    for ( i = 0 ; i < size; i++ )
        v[i] = i;
    for ( i = 0 ; i < size-1 ; i++ ) {
        j = (long int) ( ranU01( ) * (size - i));
        help = v[i];
        v[i] = v[i+j];
        v[i+j] = help;
    }
    return v; }
```